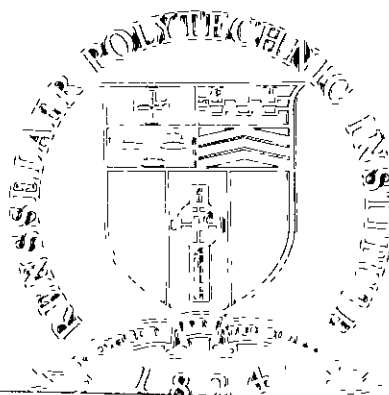
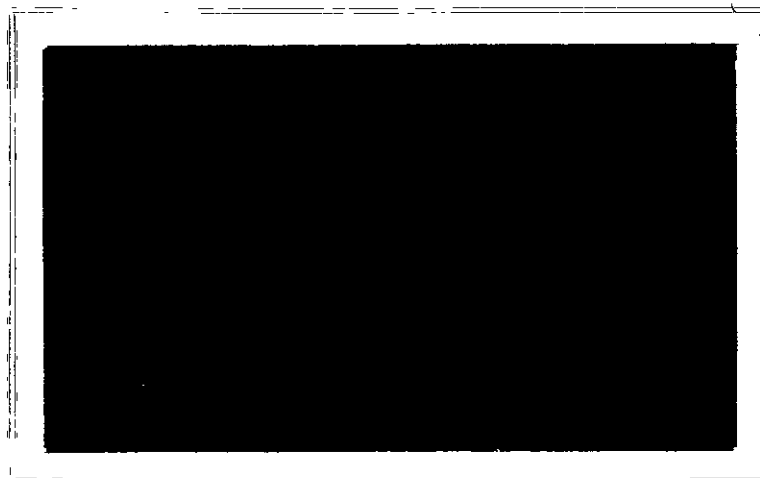


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Optimal Time Invariant Output Feedback Controllers
by Derek E. McBrinn

Submitted on behalf of
Rob Roy
Professor
Systems Engineering Division

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LIST OF SYMBOLS

<u>Symbol</u>	<u>Meaning</u>	<u>Page First Used</u>
A	System matrix	2
\hat{A}	Equivalent system matrix	36
B	Control matrix	2
C	Output/state matrix	8
\hat{C}	Equivalent output/state matrix	9
D	Output/control matrix	8
D	Arbitrary square matrix	14
D(t)	Riccati matrix	2
F	Terminal state cost matrix	2
G	Gradient matrix	15
GF	Gain function	41
I	Identity matrix	8
I_p	Permutation matrix	9
J	Cost functional	2
K	Matrix of feedback gains	3
K_0	Base gain matrix	47
\hat{K}	Perturbed gain matrix	47
M	Modal matrix	51
NC	Number of controls	2
NF	Number of feedback states	2
NS	Number of states	2
P	Equivalent cost matrix	37
Q	State cost matrix	2
$\hat{Q}, \hat{R}, \hat{W}$	Cost matrices used in Cassidy's SOC technique	7
R	Control cost matrix	2

<u>Symbol</u>	<u>Meaning</u>	<u>Page First Used</u>
T	Terminal time for optimization problem	2
V	Covariance matrix	39
W	Matrix of necessary conditions	44
t	Time	2
u	Scalar control	56
$\underline{u}(t)$	Control vector	2
$\underline{v}^T, \underline{v}_i^T$	Row eigenvector of a matrix	14
$\underline{w}, \underline{w}_i$	Column eigenvector of a matrix	14
$\underline{x}(t)$	State vector	2
$\hat{\underline{x}}(t)$	Subset of state vector	9
$\underline{y}(t)$	Output vector	8
$\underline{\bar{y}}(t)$	Alternative state vector	40
$\underline{z}(t)$	Alternative state vector	9
α	Parameter of a matrix	14
Δ	Small perturbation	16
δ	Small perturbation	47
ϵ	Element of the set	2
$\phi(t)$	State transition matrix	36
Λ	Diagonal matrix of eigenvalues	51
λ, λ_i	Eigenvalue of a matrix	14
μ	Step size (convergence) parameter	16
∇	Gradient matrix	16
(t)	Function of time	2
• as in $\dot{\underline{x}}(t)$	Differentiation with respect to time	2
T as in B^T	Transform of matrix or vector	2
$^{-1}$ as in R^{-1}	Inverse of matrix	2

l

<u>Symbol</u>	<u>Meaning</u>	<u>Page</u> <u>First</u> <u>Used</u>
\mathbf{a} as in \mathbf{a}_K	Vector arrangement of a matrix by column ordering	15
$[0, T]$	Closed time interval from 0 to T	2
i, j as in k_{ij} or $\begin{bmatrix} \\ \end{bmatrix}_{ij}$	Element in row i , column j of the matrix	10
i as in x_i	i^{th} component of vector	15
q as in b_q	q^{th} column of the matrix B	14
$\left \begin{matrix} T \\ \end{matrix} \right $	Evaluation at given point	33
$\ \ $	Euclidian norm	16
$\text{Re} \left[\begin{matrix} \\ \end{matrix} \right]$	Real part	15
$E \left[\begin{matrix} \\ \end{matrix} \right]$	Expected value	38
$\exp \left[\begin{matrix} \\ \end{matrix} \right]$	Exponentiation	37
$\text{tr} \left[\begin{matrix} \\ \end{matrix} \right]$	Trace of matrix	33

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ABSTRACT

The problem of the determination of optimal time-invariant output feedback controllers for linear dynamic systems with quadratic cost functionals is considered. Two distinct cases arise, depending on whether optimization is over a finite or a semi-infinite time interval.

For the semi-infinite time interval problem a gain initialization technique is derived to complement existing optimization techniques. The gain initialization technique determines the feedback gains required to (locally) maximize the system stability. A computational algorithm for the technique is incorporated in a digital computer program, and is used to stabilize a seven state model of a Saturn V booster rocket.

For the finite time interval problem a technique is derived to (locally) minimize the expected value of the cost functional. The technique uses the concept of Initial State Averaging. A computational algorithm is provided and incorporated in a digital computer program. The technique is illustrated by three examples.

CHAPTER I

INTRODUCTION

I.1 Motivation

During the past decade much excellent work has been done in the field of Optimal Control Theory. Many books and a great number of technical papers have been published - the Bibliography of this report cites only a small portion of the total. Not all of this work has been purely theoretical. A substantial background has been developed in the important practical area of computation and it has for several years been feasible to determine optimal controllers for a wide variety of physical systems.

It would be expected that practical applications of this theory would have appeared in abundance. Optimization is the essence of all good engineering design, and practicing engineers should seize upon any techniques which might aid them in their work. Application of the new techniques has, however, been disappointingly slow. With a few notable exceptions the practical design of control systems has remained based on the transfer-function techniques developed prior to the 1960's.

The explanation of the above paradox is widely recognized. It lies in the nature of the word "optimal", which is meaningless without a criterion of optimality. By and large, control theorists have used criteria of optimality dependent only on the performance of the control system. Design engineers, on the other hand, interpret "optimal" as embracing both system performance and system cost. The difference may be best illustrated by a simple example.

Consider the problem of designing a speed control device for a cheap movie camera. The objective is to ensure that sixteen frames of film are

exposed per second regardless of film tension, state of charge of the motor batteries, etc. This could be interpreted as comprising an example of the state-regulator problem in Optimal Control Theory, with the following method of analysis. The equations governing the film transportation are determined and linearized to the form

$$\dot{\underline{x}}(t) = A \underline{x}(t) + B \underline{u}(t) \quad (I.1-1)$$

where $\underline{x}(t)$ is an NS-vector describing the deviation of the system from its desired state at time t

$\underline{u}(t)$ is an NC-vector defining the control inputs at time t

A and B are matrices whose elements depend on the characteristics of the system.

A cost functional is formulated having the form

$$J = \underline{x}^T(T) F \underline{x}(T) + \int_0^T [\underline{x}^T(\tau) Q \underline{x}(\tau) + \underline{u}^T(\tau) R \underline{u}(\tau)] d\tau \quad (I.1-2)$$

where T is the duration of the scene to be filmed. The objective is to determine the control input $\underline{u}(t)$ for $t \in [0, T]$ which minimizes the cost functional J . The positive semi-definite matrices F and Q are chosen to penalize any given control system for the deviations it allows from the desired state. The positive definite matrix R is a penalty for improvident usage of control power (i.e., battery energy). The solution to the problem defined by (I.1-1) and (I.1-2) can be shown² to be

$$\underline{u}(t) = -R^{-1} B^T D(t) \underline{x}(t) \quad (I.1-3)$$

where $D(t)$ satisfies the matrix Riccati equation

$$\dot{D}(t) = -D(t) A - A^T D(t) + D(t) B R^{-1} B^T D(t) - Q \quad (I.1-4)$$

with

$$D(T) = F \quad (I.1-5)$$

Defining

$$K^T(t) = R^{-1} B^T D(t) \quad (I.1-6)$$

and re-writing (I.1-3) as

$$\underline{u}(t) = -K^T(t) \underline{x}(t) \quad (I.1-7)$$

we see that the optimal control input can be generated as a time-varying linear combination of all of the states of the system. We note in passing that the less-fastidious photographer, willing to edit a few frames from the beginning and end of each scene and settle for good steady state performance, would be rewarded by simplification of the controller to the form²

$$\underline{u}(t) = -K^T \underline{x}(t) \quad (I.1-8)$$

This time-invariant controller would be much simpler to mechanize than that described by (I.1-7).

Now consider that speed control of the typical cheap movie camera is achieved by means of a simple flyball governor driving an on/off switch between the batteries and the motor. The flyball governor, invented by James Watt in 1788, was the first widely used automatic feedback controller.⁵ In its simpler form it suffers from "hunting" about the set point. In its application to the movie camera it is driven by the single system state, motor speed. The single feedback gain can be considered as constant in the vicinity of the set point. It is clear that the flyball governor does not satisfy the requirements for an optimal controller as defined by (I.1-7).

Yet the flyball governor is indeed an optimal controller in the best engineering sense. It does an acceptable job at minimum cost. The mechanization of the "optimal" controller defined by (I.1-7) on the other hand, would be prohibitively expensive. It would require measurement and feedback of all system states. It would also require a digital computer to calculate and store the time-varying feedback gains for each scene to be filmed.

It should be clear from the above that the allowable degree of complexity of a control system is often constrained by considerations of cost. Weight, reliability and common engineering sense similarly often dictate simplicity. What the control system designer wants then is not that system which performs in the best possible manner. Rather he wishes the best possible performance for a given degree of complexity. This problem is considerably more difficult than that defined by (I.1-1) and (I.1-2). It must however, be solved if the benefits of optimal control are ever to be realized for the majority of potential applications.

This report considers a portion of the general problem of optimization within a given degree of control system complexity. Only linear systems with quadratic cost functionals are considered. Such combinations may be described by (I.1-1) and (I.1-2). The allowable degree of complexity of the controller is assumed to be time-invariant feedback of the system outputs only. The system outputs are those quantities which can be measured, and in most cases provide less than a full description of the system state at any time. The above limitations are those which are normally applied in classical control system analysis techniques. Thus this report seeks to optimize the classical controller, without increasing its complexity.

I.2 Historical Review

It is generally conceded that control system design prior to the Second World War was primarily an art. The techniques used were often empirical and thus confined to particular classes of problems. This situation was largely eliminated by the contributions of Nyquist,⁶ Bode⁷ and Evans⁸ who established the theory and techniques of control system design which predominate in practical work even today. These techniques, however, are most suited to the design of relatively simple systems. They work best for systems having a single input and a single output related by a transfer function. Also these techniques are purely analytical and cannot be used to directly synthesize a control system.

The problem of synthesis was first tackled by Wiener⁹, who considered the optimization of linear filters. This work was extended to control systems by Newton, Gould and Kaiser¹⁰, who, however, still retained the transfer-function approach; their technique was to determine the values of system parameters required to minimize a cost functional for a specific perturbation of the system. It is significant that they made use of the Calculus of Variations in their approach to this problem. The extension to the classical Calculus of Variations provided by Pontryagin's Maximum Principle,¹¹ and the control problem framework provided by Kalman¹² finally brought Optimal Control Theory to maturity. This maturity, combined with the state-variable formulation presented by DeRusso, Roy and Close¹ and the increasing capability of digital computers, finally allowed the determination of optimal controllers for a wide variety of systems.

The early excitement at the disclosure of the ability to compute optimal controllers faded when it was realized that such controllers were very difficult to mechanize. The reasons for this difficulty are twofold. First, an optimal controller requires knowledge of the complete state of

the system. It may not, however, be feasible to measure all system states. Secondly, for finite time duration problems an optimal controller requires time-varying feedback.

There are two fundamental approaches to elimination of the first difficulty. The approach taken by Kalman and Bucy¹³ was to estimate the unmeasurable states from the (noise corrupted) system outputs or measurable states. This approach was extended by Luenberger¹⁴ and again by Ash¹⁵. Estimation of the unmeasurable states allows the mechanization of a regular optimal controller. Note, however, that it adds the complexities of a state estimator to those of the optimal controller.

The second approach to the problem of unmeasurable states is to design a controller which uses only the available information. Such a controller will be "sub-optimal" in the mathematical sense of minimizing a cost functional such as that described by (I.1-2). It may well, however, be optimal in the engineering sense. This approach has been tried by a number of people. Newton, Gould and Kaiser¹⁰ essentially advocated such controllers, but optimized them for specific system disturbances. Consideration of only a specific disturbance is equivalent to imposing a specific initial condition on the system to be controlled. Max-min techniques were developed in an attempt to eliminate the dependence of the solution on the chosen initial condition.^{16,17} In the max-min procedure the maximum system cost with respect to a set of initial conditions is minimized with respect to the system feedback gains. It still remains to choose an appropriate set of initial conditions. Rekasius¹⁸ circumvented this problem by considering the initial condition giving the maximum ratio of output feedback cost to allstate feedback cost but was still limited to design for this specific initial condition. Levine¹⁹ determined the

controller which was optimal in an average sense and computed time varying feedback gains for the case of the finite time-interval problem. For the semi-infinite time interval problem he required his feedback gains to be time-invariant. Cassidy²⁰ considered the minimization of a modified cost functional having the form

$$J = \frac{1}{2} \int_0^T (\underline{x}^T \underline{Q} \underline{x} + \underline{x}^T \hat{\underline{Q}} \underline{x} + \underline{x}^T \underline{W} \underline{u} + \underline{x}^T \hat{\underline{W}} \underline{y} + \underline{u}^T \underline{R} \underline{u}) d\tau \quad (1.2-1)$$

where the matrices $\hat{\underline{Q}}$ and $\hat{\underline{W}}$ were chosen to ensure that the feedback gains for the unavailable states were zero. The resultant feedback gains were again time-varying for a finite time interval, time-invariant for the semi-infinite time interval.

It is considered that the approaches used by Levine and Cassidy are the most satisfactory of those considered, since they are both completely independent of initial conditions. They both, however, result in time-varying gains for the finite time-interval case. For the semi-infinite time interval both give constant feedback gains, but use computational algorithms which require initialization by a set of stabilizing feedback gains.

The problem of determining stabilizing output-feedback gains for complex systems has not, to the author's knowledge, been satisfactorily solved. Koenigsberg²¹ has considered local stability maxima, but his computational algorithm is felt to be inefficient. Jameson²² and Davison²³ have shown that as many system roots as there are feedback states can be arbitrarily determined, but the remaining roots are then unconstrained.

A solution to the problem of determining optimal constant or piecewise constant feedback gains for a finite time interval has been derived by

Kleinman, Fortmann and Athans.²⁵ As with Levine,¹⁹ they considered optimality in an average sense, but confined attention to the allstate feedback case.

I.3 Scope and Contribution of This Work

This report considers the determination of optimal time-invariant output feedback control gains for systems which can be described by the equation

$$\dot{\underline{x}}(t) = A \underline{x}(t) + B \underline{u}(t) \quad (\text{I.3-1})$$

where $\underline{x}(t)$ is an NS-vector describing the state of the system at time t
 $\underline{u}(t)$ is an NC-vector describing the control inputs at time t
 A, B are time invariant matrices of appropriate order.

It is assumed, without loss of generality, that the system outputs are the first NF states. If this is not the case then a new set of state variables may be obtained as follows to satisfy the assumption. Suppose that the system described by (I.3-1) has NF independent outputs $\underline{y}(t)$ given by

$$\underline{y}(t) = C \underline{x}(t) + D \underline{u}(t) \quad (\text{I.3-2})$$

We shall require that

$$\underline{u}(t) = -K^T \underline{y}(t) \quad (\text{I.3-3})$$

where K is the time-invariant matrix of feedback gains. Thus

$$\left[I + DK^T \right] \underline{y}(t) = C \underline{x}(t) \quad (\text{I.3-4})$$

where I is the identity matrix of order NF. Assuming $\left[I + DK^T \right]$ to be non-singular gives

$$\underline{y}(t) = \hat{C} \underline{x}(t) \quad (I.3-5)$$

where

$$\hat{C} = [I + DK^T]^{-1} C \quad (I.3-6)$$

We note that in most cases D will be zero, leaving \hat{C} equal to C . We now form the vector $\hat{\underline{x}}(t)$ comprising any $(NS - NF)$ of the $\underline{x}(t)$ state variables which are independent of the $\underline{y}(t)$ variables. We note that such variables must exist since $\underline{x}(t)$ spans NS space while $\underline{y}(t)$ only spans NF space. Appending $\hat{\underline{x}}(t)$ to $\underline{y}(t)$ gives

$$\underline{z}(t) = \begin{bmatrix} \underline{y}(t) \\ \hat{\underline{x}}(t) \end{bmatrix} = \begin{bmatrix} \hat{C} \\ I_p \end{bmatrix} \underline{x}(t) \quad (I.3-7)$$

where I_p is a permutation matrix describing how $\hat{\underline{x}}(t)$ was chosen from $\underline{x}(t)$.

* It follows that

$$\dot{\underline{z}}(t) = \begin{bmatrix} \hat{C} \\ I_p \end{bmatrix} \dot{\underline{x}}(t) \quad (I.3-8)$$

i.e.,

$$\dot{\underline{z}}(t) = \begin{bmatrix} C \\ I_p \end{bmatrix} A \underline{x}(t) + \begin{bmatrix} C \\ I_p \end{bmatrix} B \underline{u}(t) \quad (I.3-9)$$

Thus, from (I.3-7)

$$\dot{\underline{z}}(t) = \begin{bmatrix} C \\ I_p \end{bmatrix} A \begin{bmatrix} C \\ I_p \end{bmatrix}^{-1} \underline{z}(t) + \begin{bmatrix} C \\ I_p \end{bmatrix} B \underline{u}(t) \quad (I.3-10)$$

and $\underline{z}(t)$ is the desired state-vector having its first NF states as the system outputs.

Thus the control inputs $\underline{u}(t)$ for the system described by (I.3-1) are to be optimized over the set

$$\underline{u}(t) = -K^T \underline{x}(t) \quad (\text{I.3-11})$$

where K , the time invariant matrix of feedback gains, is constrained to the form

$$K = \left[\begin{array}{c|c} \overbrace{\begin{matrix} k_{ij} \\ \vdots \\ \vdots \end{matrix}}^{NC} & \left. \begin{matrix} \\ \vdots \\ \vdots \end{matrix} \right\}^{NF} \right] \left. \begin{matrix} \\ \vdots \\ 0 \end{matrix} \right\}^{NS} \quad (\text{I.3-12})$$

The cost functional is assumed to be of the form

$$J = \underline{x}^T(T) F \underline{x}(T) + \int_0^T [\underline{x}^T(\tau) Q \underline{x}(\tau) + \underline{u}^T(\tau) R \underline{u}(\tau)] d\tau \quad (\text{I.3-13})$$

where F and Q are positive semi-definite

R is positive definite

and optimization is over the time interval $[0, T]$.

Two separate cases are investigated. For the case where the terminal time T is infinite it is considered that the techniques presented by either Levine¹⁹ or Cassidy²⁰ are adequate. Both these techniques, however, require initialization by stabilizing feedback gains. In practice, indeed the author has found that feedback gains giving marginal stability may be inadequate, due to numerical considerations. Chapter II of this report therefore presents a computational algorithm designed to find feedback gains of the form given by (I.3-12) such that the system stability is driven to a local maximum. Note that maximum stability is here defined to mean

that the least stable of the system roots is made as stable as possible. The algorithm presented is felt to be superior to that given by Koenigsberg²¹ in that it is computationally faster.

For the case where the terminal time T is finite, an approach is taken similar to that used by Levine¹⁹, except that the feedback gains are required to be time-invariant. The resulting theory is presented in Chapter III, together with a computational algorithm to allow the determination of the optimal gains.

It is felt that this report will help to fill some of the gaps in the existing ability to determine practical feedback controllers which are optimal in the engineering sense.

CHAPTER II

THE GAIN INITIALIZATION PROBLEM

II.1 Introduction

The problem of determination of the output feedback gains required to stabilize a dynamic system is an interesting one in its own right. In this report, however, we are concerned with the determination of such gains as a pre-requisite to the application of optimization techniques.

Both Levine¹⁹ and Cassidy²⁰ have considered the determination of optimal output feedback controllers for linear systems. Both have derived iterative techniques resulting in time invariant feedback gains when the system is optimized over the semi-infinite time interval $[0, \infty]$. Each technique, however, requires initialization with a set of stabilizing feedback gains in order to ensure convergence of the computational algorithm.

Based on the author's practical experience²⁶ with Cassidy's technique it appears that an extra requirement on initial system stability may be added when the optimization algorithms are mechanized on a digital computer. It was found that with low order systems (five states or less) marginal stability was sufficient to ensure convergence of the optimization algorithm. When a twenty-one state model of the Saturn V booster rocket was considered, however, initialization by feedback gains giving marginal stability was insufficient to cause convergence. This was judged to be due to numerical truncation resulting from the finite word length in the digital computer. The problem was solved by computing a new set of feedback gains giving better than marginal stability. Re-initialization of the optimization algorithm with the new feedback gains resulted in convergence. It is considered likely that Levine's algorithm would have similar characteristics.

It is clear that practical application of Levine's and Cassidy's techniques requires a method for determination of stabilizing output feedback gains. It would be desirable that the gains give more than marginal stability. A method is described in this Chapter for determination of such gains, when they exist.

II.2 Theory of the Gain Initialization Technique

The problem under consideration is as follows. Given the system

$$\dot{\underline{x}}(t) = A \underline{x}(t) + B \underline{u}(t) \quad (\text{II.2-1})$$

where $\underline{x}(t)$ is an NS-vector describing the state of the system at time t

$\underline{u}(t)$ is an NC-vector describing the control inputs at time t

A, B are matrices whose elements depend on the characteristics of the system

with

$$\underline{u}(t) = -K^T \underline{x}(t) \quad (\text{II.2-2})$$

where K is a matrix of feedback gains such that

$$\dot{\underline{x}}(t) = \left[A - BK^T \right] \underline{x}(t) \quad (\text{II.2-3})$$

where K is constrained to the form

$$K = \left\{ \begin{array}{c} \text{NC} \\ \left[\begin{array}{c} k_{ij} \\ \dots \\ 0 \end{array} \right] \end{array} \right\} \text{NF} \left. \vphantom{\begin{array}{c} \text{NC} \\ \left[\begin{array}{c} k_{ij} \\ \dots \\ 0 \end{array} \right] \end{array}} \right\} \text{NS} \quad (\text{II.2-4})$$

how should the feedback gain matrix K be modified, within the constraints of (II.2-4) to increase the stability of the least stable eigenvalue of (II.2-3)

The solution hinges on an expression given by Fadeev and Fadeeva²⁷ for the sensitivity of an eigenvalue of a matrix to a parameter of the matrix. Given any eigenvalue λ and corresponding row and column eigenvectors \underline{v}^T and \underline{w} of a square matrix D , the sensitivity of the eigenvalue to a parameter α of the matrix is given by

$$\frac{d\lambda}{d\alpha} = \frac{\underline{v}^T \frac{\partial D}{\partial \alpha} \underline{w}}{\underline{v}^T \underline{w}} \quad (\text{II.2-5})$$

We note that \underline{v} and \underline{w} satisfy

$$D^T \underline{v} = \lambda \underline{v} \quad (\text{II.2-6})$$

$$D \underline{w} = \lambda \underline{w} \quad (\text{II.2-7})$$

To apply (II.2-5) to the system described by (II.2-3) and (II.2-4) we note that

$$\frac{\partial [A - BK^T]}{\partial k_{pq}} = \underbrace{\begin{bmatrix} 0 & \vdots & \underline{b}_q & \vdots & 0 \end{bmatrix}}_{p^{\text{th}} \text{ column}} \quad (\text{II.2-8})$$

where \underline{b}_q is the q^{th} column of the matrix B . Thus if λ_i is any eigenvalue of (II.2-3) and if \underline{v}_i^T and \underline{w}_i are corresponding row and column eigenvectors we see that

$$\frac{\partial \lambda_i}{\partial k_{pq}} = \frac{\underline{v}_i^T \underbrace{\begin{bmatrix} 0 & \vdots & \underline{b}_q & \vdots & 0 \end{bmatrix}}_{p^{\text{th}} \text{ column}} \underline{w}_i}{\underline{v}_i^T \underline{w}_i} \quad (\text{II.2-9})$$

The sensitivity of the real part of λ_i to the real feedback gain k_{pq} is simply the real part of the right hand side of (II.2-9).

Suppose now that it is desired to decrease the real part of λ_i . The direction of the steepest descent in feedback gain space is defined by the gradient matrix G where

$$g_{pq} = \operatorname{Re} \left\{ \frac{\partial \lambda_i}{\partial k_{pq}} \right\} \quad (\text{II.2-10})$$

$$p = 1, \dots, NF$$

$$q = 1, \dots, NC$$

II.3 The Gain Initialization Algorithm

A gain initialization algorithm has been derived from the expressions given in (II.2-9) and (II.2-10), and has been incorporated in the digital computer program GRADGN. A listing of the program GRADGN is given in Appendix I. A description of the algorithm/program follows.

The algorithm is iterative. It increases the stability of the least stable eigenvalue of the system at each step. If, during a given iteration a new eigenvalue should become less stable than the one being operated on, this fact is taken into account in subsequent iterations.

Consider the rearrangement of the variable elements of the feedback gain matrix K into an $NF \times NC$ vector ${}^n K^n$. The rearrangement is performed by column ordering of K so that

$$K_{pq} = {}^n K^n_{(q-1)NF + p} \quad (\text{II.3-1})$$

The gradient matrix G is similarly rearranged giving

$$G_{pq} = {}^n G^n_{(q-1)NF + p} \quad (\text{II.3-2})$$

The steepest descent method requires that variations in the gain vector \mathbf{K}^a must lie along the negative of the gradient vector \mathbf{G}^a .

Thus

$$\mathbf{\Delta K}^a = -\mu \mathbf{G}^a \quad (\text{II.3-3})$$

where μ is a positive constant determined by the step size to be taken and $\mathbf{\Delta K}^a$ is a variation in the gain vector. Suppose now that we wish to determine the variation $\mathbf{\Delta K}^a$ in the gain vector \mathbf{K}^a required to cause a small negative change $\Delta \text{Re}\{\lambda_i\}$ in the real part of the eigenvalue λ_i . The variation $\mathbf{\Delta K}^a$ may be computed from the relationship

$$\mathbf{G}^{aT} \mathbf{\Delta K}^a = \Delta \text{Re}\{\lambda_i\} \quad (\text{II.3-4})$$

When the variation $\mathbf{\Delta K}^a$ lies in the direction of steepest descent then, from (II.3-3) and (II.3-4)

$$-\mathbf{G}^{aT} \mu \mathbf{G}^a = \Delta \text{Re}\{\lambda_i\} \quad (\text{II.3-5})$$

whence

$$\mu = \frac{-\Delta \text{Re}\{\lambda_i\}}{\|\mathbf{G}^a\|^2} \quad (\text{II.3-6})$$

and so

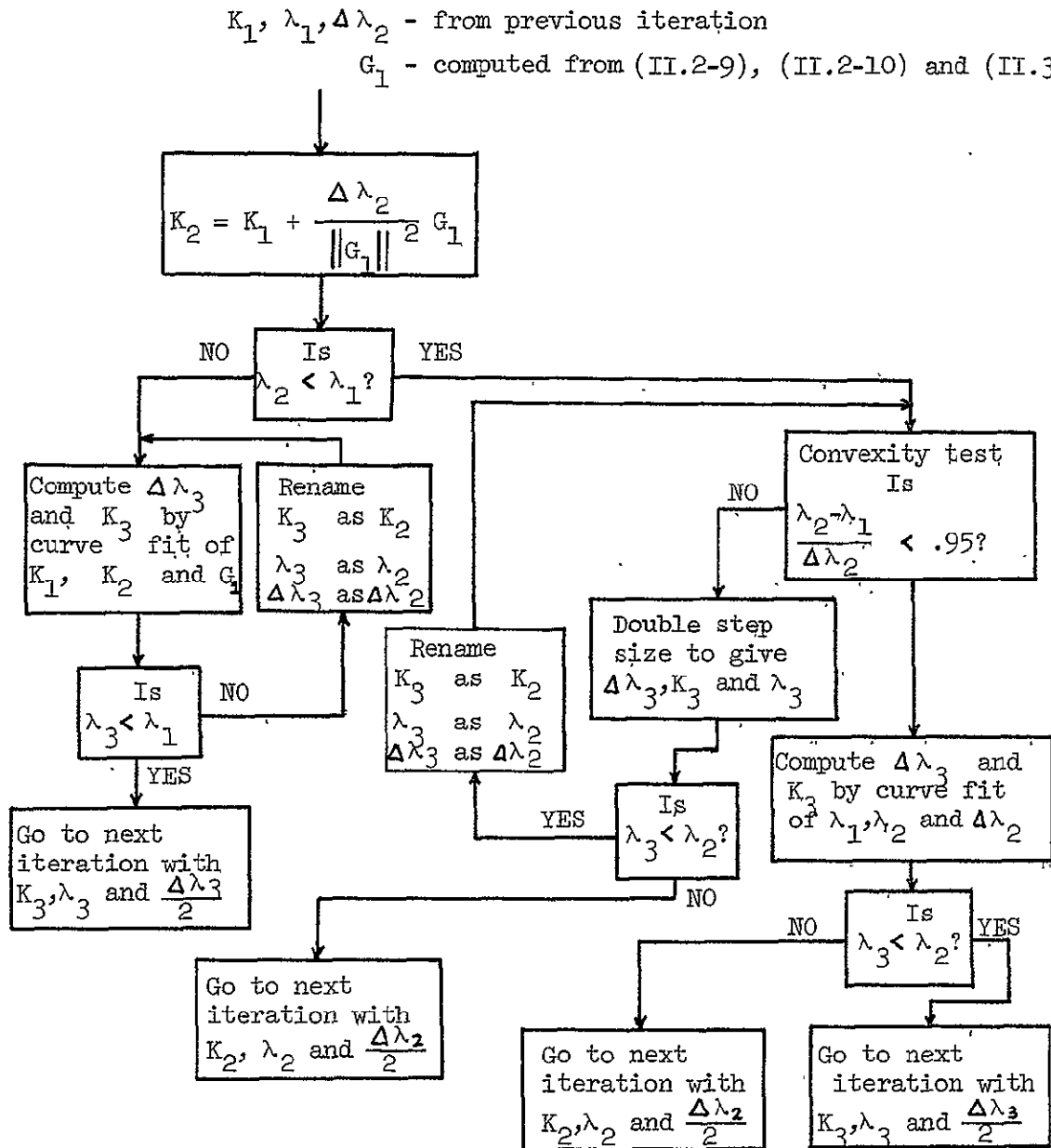
$$\mathbf{\Delta K}^a = \frac{-\Delta \text{Re}\{\lambda_i\}}{\|\mathbf{G}^a\|^2} \mathbf{G}^a \quad (\text{II.3-7})$$

When the variation $\Delta \text{Re}\{\lambda_i\}$ is sufficiently small the system corresponding to the feedback gains $(\mathbf{K}^a + \mathbf{\Delta K}^a)$ should have an eigenvalue whose real part approximates $\text{Re}\{\lambda_i\} + \Delta \text{Re}\{\lambda_i\}$.

It remains to determine suitable values for the step size $\Delta \text{Re} \{ \lambda_i \}$. It is assumed that in most cases a user of GRADGN will have little idea of what comprises a suitable step size. The step size is therefore varied adaptively based on experience with previous iterations. The initial step size may either be assigned by the user or given a default value of -0.1. In either case the initial step size is used as a program criterion for termination due to diminishing returns. Should two successive iterations of the program fail to improve the stability by the initial step size, the program is terminated. It should be noted that apart from its use as a termination control the program is quite insensitive to the initial step size, due to the rapidity of the step size adaptation.

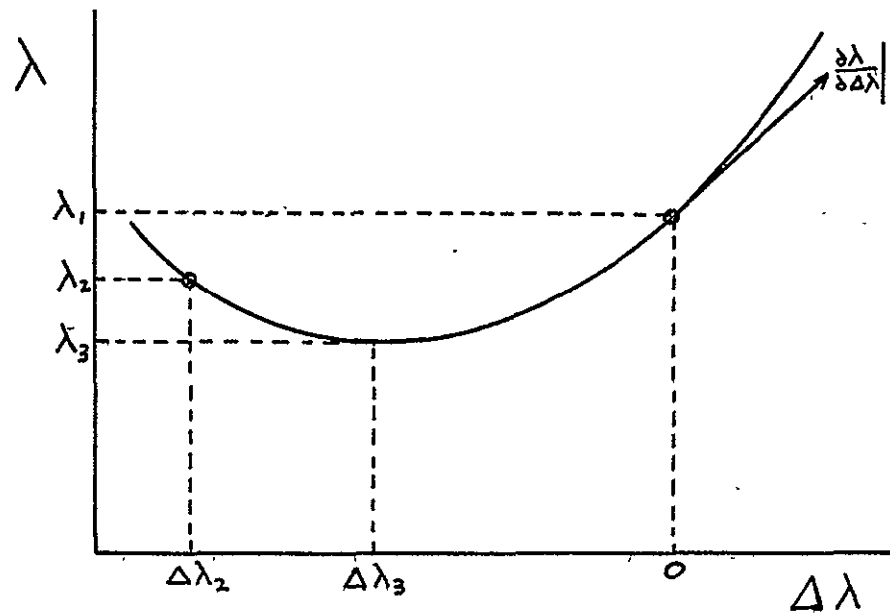
The adaptation control for a given iterative step is described by Figure II.3-1. One of two basic step size adjustments is utilized, depending on the convexity of the trajectory of the least stable eigenvalue. If the trajectory is sufficiently convex then a quadratic curve is fitted to two points on the trajectory and the trajectory gradient at one of the points. This procedure is illustrated by Figure II.3-2. The gain vector corresponding to the minimum point of the quadratic curve is computed and tested for stability. If the trajectory is insufficiently convex then the step size is doubled. One or both of these procedures may be used repeatedly during a single iteration. Note that only one gradient vector is computed per iteration.

The initiation point for the next iteration is always the most stable gain vector found. The starting step size for the next iteration is set to one half of the current improvement in the least stable eigenvalue. This allows the step size to be changed by several orders of the magnitude in either direction during a single iteration, while allowing for the decreasing rate of improvement as a stability maximum is approached.



K_i = feedback gain vector at computation point i
 λ_i = real part of least stable eigenvalue corresponding to feedback gains K_i
 G_1 = gradient vector at computation point 1
 $\Delta\lambda_i$ = step size from computation point 1 to computation point i

FIGURE II.3-1 Step Size Adaptation Control for the
 Gain Initialization Program GRADGN



$\Delta \lambda_i$ = step size from computation point 1 to computation point i

λ_i = real part of least stable eigenvalue at computation point i

Quadratic curve fit to variation of λ with $\Delta \lambda$ gives

$$\lambda = a(\Delta \lambda)^2 + b \Delta \lambda + c$$

and

$$c = \lambda_1$$

$$b = \left. \frac{\partial \lambda}{\partial \Delta \lambda} \right|_1 = 1$$

$$a = \frac{\lambda_2 - b \Delta \lambda_2 - c}{(\Delta \lambda_2)^2}$$

Curve has minimum point at $\Delta \lambda_3$ where $\left. \frac{\partial \lambda}{\partial \Delta \lambda} \right|_3 = 0$

$$\therefore \Delta \lambda_3 = -\frac{b}{2a} = -\frac{(\Delta \lambda_2)^2}{2(\lambda_2 - \lambda_1 - \Delta \lambda_2)}$$

FIGURE II.3-2 Step Size Adaptation in GRADGN by Quadratic Curve Fit

The computer program GRADGN incorporates an option for the degree of stability required for termination. Setting the input parameter ISTOP to zero instructs the program to maximize stability. Termination is then initiated by reaching a point of diminishing returns, as mentioned above. If ISTOP is set of 1 a minimum desired degree of stability, STOPR, may be input to the program. The program then terminates when the real part of the least stable eigenvalue becomes less than STOPR.

II.4 Example of the Use of the Gain Initialization Program GRADGN

The use of the gain initialization program GRADGN is illustrated in this Section by output feedback stabilization of a seven state model of the Saturn V booster rocket. The model represents the rocket dynamics at a point occuring 80 seconds after lift-off.

The seven states considered are

$$\begin{aligned}
 x_1 &= \text{measured pitch attitude angle } \phi_D \\
 x_2 &= \text{measured pitch rate } \dot{\phi}_R \\
 x_3 &= \text{aerodynamic angle of attack } \alpha \\
 x_4 &= \text{first bending mode deflection } \eta_1 \\
 x_5 &= \text{first bending mode deflection rate } \dot{\eta}_1 \\
 x_6 &= \text{engine gimbal angle } \beta \\
 x_7 &= \text{engine gimbal angle rate } \dot{\beta}
 \end{aligned}$$

It is assumed that only the first two states x_1 and x_2 are to be measured. Thus only these two states are available for feedback. Control is achieved via an actuator driving the engine gimbal angle at a rate proportional to the actuator input.

The model may be represented by

$$\dot{\underline{x}}(t) = A \underline{x}(t) + \underline{b} \underline{u}(t) \quad (\text{II.4-1})$$

with

$$\underline{u}(t) = -k_1 x_1(t) - k_2 x_2(t) \quad (\text{II.4-2})$$

where

$$A = \begin{bmatrix} 0. & 1. & 0. & 0. & 0. & 0. & 0. \\ 0. & 0. & .2030 & -.6535 & -.0020 & 2.558 & 0. \\ -.0137 & 1. & -.0407 & .0002 & -.0146 & -.0334 & 0. \\ 0. & 0. & 0. & 0. & 1. & 0. & 0. \\ 0. & 0. & 0. & -44.67 & -.1337 & 254.6 & 0. \\ 0. & 0. & 0. & 0. & 0. & 0. & 1. \\ 0. & 0. & 0. & 0. & 0. & -50. & -10. \end{bmatrix} \quad (\text{II.4-3})$$

$$\underline{b} = \begin{bmatrix} 0. \\ 0. \\ 0. \\ 0. \\ 0. \\ 0. \\ 1. \end{bmatrix} \quad (\text{II.4-5})$$

The derivation of this model is described by Cassidy.²⁰

The program GRADGN was used to determine the feedback gains k_1 and k_2 corresponding to a local stability maximum of the above system. The program output is given in Figure II.4-1.

STATES	CONTROLS	FEEDBACKS	GAINS	IDEAL	ISTOP
7	1	2	0	1	0
STABILITY INCREASE INCREMENT			0.04000000		
MINIMUM STABILITY REQUIRED FOR TERMINATION			-1000000.00000000		
SYSTEM MATRIX A MAT					
0.0000000E 00	0.1000000E 01	0.0000000E 00	0.0000000E 00	0.0000000E 00	
0.0000000E 00	0.0000000E 00	0.0000000E 00	0.0000000E 00	0.0000000E 00	
0.0000000E 00	0.2030470E 00	-0.6534950E 00	-0.1955000E 02	0.0000000E 00	
0.2558020E 01	0.0000000E 00	0.1366150E 01	0.0000000E 00	0.0000000E 00	
0.4068250E 01	0.1998600E 03	-0.1463000E 01	-0.3338200E 01	0.0000000E 00	
0.0000000E 00	0.0000000E 00	0.0000000E 00	0.0000000E 00	0.0000000E 00	
0.0000000E 00	0.1000000E 01	0.0000000E 00	0.0000000E 00	0.0000000E 00	
0.0000000E 00	0.0000000E 00	0.0000000E 00	0.0000000E 00	0.4466811E 02	
0.1336680E 00	0.2546100E 03	0.0000000E 00	0.0000000E 00	0.0000000E 00	
0.0000000E 00	0.0000000E 00	0.0000000E 00	0.0000000E 00	0.0000000E 00	
0.0000000E 00	0.1000000E 01	0.0000000E 00	0.0000000E 00	0.0000000E 00	
0.0000000E 00	0.0000000E 00	0.0000000E 00	0.0000000E 00	-0.5000000E 02	
0.1000000E 02					
CONTROL MATRIX B MAT					
0.0000000E 00	0.0000000E 00	0.0000000E 00	0.0000000E 00	0.0000000E 00	
0.0000000E 00	0.0000000E 00	0.0000000E 00	0.1000000E 01	0.0000000E 00	
GAIN MATRIX K					
0.0000000E 00	0.0000000E 00				
ROOTS					
REAL PART		IMAG. PART			
-0.47798E 00		0.000000E 00			
0.423402E 00		0.000000E 00			
0.137118E 01		0.000000E 00			
0.668340E 01		0.668309E 01			
0.668340E 01		-0.668309E 01			
0.500000E 01		0.500000E 01			
0.500000E 01		-0.500000E 01			
MAXIMUM REAL PART OF ROOTS					
0.423402E 00					

NOT REPRODUCIBLE

Figure II.4-1 Computer Print-Out For
Example Problem

GAIN SEARCH PROCEEDS ALONG A NEW GRADIENT. ITERATION NUMBER 1			
EIGVEC ERROR MESSAGES			
SWI=0.8345E-06		ITER= 4	DIF=0.3576E-06
EIGENVECTORS CORRESP TO MRP EIGENVALUE:			
ROW REAL PART	ROW IMAG PART	COL REAL PART	COL IMAG PART
-0.1411699E-01	0.0000000E 00	0.1000000E 01	0.0000000E 00
0.1000000E 01	0.0000000E 00	0.4234034E 00	0.0000000E 00
0.4375202E 00	0.0000000E 00	0.8829014E 00	0.0000000E 00
0.2059637E-03	0.0000000E 00	-0.2525228E-26	0.0000000E 00
-0.1463000E-01	0.0000000E 00	0.8821868E-26	0.0000000E 00
-0.2263338E 00	0.0000000E 00	-0.1349601E-27	0.0000000E 00
-0.2171400E-01	0.0000000E 00	-0.4707256E-27	0.0000000E 00
GRADIENT MATRIX			
0.2729351E-01		0.1155617E-01	
STABILITY INCREASE STEP SIZE = 0.04000000			
GAIN MATRIX K			
-0.1242759E 01		-0.5261886E 00	
ROOTS			
REAL PART		IMAG. PART	
-0.708895E-01		0.666810E 01	
-0.708895E-01		-0.666810E 01	
-0.499273E 01		0.501011E 01	
-0.499273E 01		-0.501011E 01	
-0.452182E 00		0.000000E 00	
0.380997E 00		0.000000E 00	
0.239837E-01		0.000000E 00	
STEP SIZE IS DOUBLED			
GAIN MATRIX K			
-0.2485518E 01		-0.1052377E 01	
ROOTS			
REAL PART		IMAG. PART	
-0.747366E-01		0.665300E 01	
-0.747366E-01		-0.665300E 01	
-0.498541E 01		0.502038E 01	
-0.498541E 01		-0.502038E 01	
-0.424900E 00		0.000000E 00	
-0.332093E 00		0.000000E 00	
-0.389951E 01		0.000000E 00	

FIGURE II.4-1 (Cont'd)

STEP SIZE IS DOUBLED

GAIN MATRIX K

-0.4971038E-01 -0.2104753E-01

ROOTS	REAL PART	IMAG. PART
-0.827146E-01	0.662188E-01	
-0.827146E-01	-0.662188E-01	
-0.497082E-01	0.504117E-01	
-0.497082E-01	-0.504117E-01	
-0.364258E-00	0.000000E-00	
0.148511E-00	0.256780E-01	
0.148511E-00	-0.256780E-01	

STEP SIZE IS DOUBLED

GAIN MATRIX K

-0.9942076E-01 -0.4209508E-01

ROOTS	REAL PART	IMAG. PART
-0.979548E-01	0.655847E-01	
-0.979548E-01	-0.655847E-01	
-0.494167E-01	0.508360E-01	
-0.494167E-01	-0.508360E-01	
0.625687E-01	0.243981E-00	
0.625687E-01	-0.243981E-00	
-0.220044E-00	0.000000E-00	

STEP SIZE IS DOUBLED

GAIN MATRIX K

-0.1988414E-02 -0.8419017E-01

ROOTS	REAL PART	IMAG. PART
-0.126488E-00	0.642529E-01	
-0.126488E-00	-0.642529E-01	
-0.488370E-01	0.517278E-01	
-0.488370E-01	-0.517278E-01	
-0.315896E-01	0.533504E-00	
-0.315896E-01	-0.533504E-00	
-0.907448E-01	0.000000E-00	

STEP SIZE IS DOUBLED

FIGURE II.4-1 (Cont'd)

GAIN MATRIX K
 -0.3976830E 02 -0.1683803E 02

ROOTS	REAL PART	IMAG. PART
-0.477230E 01	0.536959E 01	
-0.477230E 01	-0.536959E 01	
-0.169915E 00	0.613263E 01	
-0.169915E 00	-0.613263E 01	
-0.113802E 00	0.913033E 00	
-0.113802E 00	-0.913033E 00	
-0.621070E -01	0.000000E 00	

STEP SIZE IS DOUBLED

ROOTS	REAL PART	IMAG. PART
-0.460448E 01	0.583911E 01	
-0.460448E 01	-0.583911E 01	
-0.144409E 00	0.542965E 01	
-0.144409E 00	-0.542965E 01	
-0.311907E 00	0.148304E 01	
-0.311907E 00	-0.148304E 01	
-0.528619E -01	0.000000E 00	

STEP SIZE TOO LARGE. COMPUTE NEW GRADIENT AT PREVIOUS BEST GAINS

ROOTS	REAL PART	IMAG. PART
-0.477230E 01	0.536959E 01	
-0.477230E 01	-0.536959E 01	
-0.169915E 00	0.613263E 01	
-0.169915E 00	-0.613263E 01	
-0.113802E 00	0.913033E 00	
-0.113802E 00	-0.913033E 00	
-0.621070E -01	0.000000E 00	

MAXIMUM REAL PART OF ROOTS
-0.621070E -01

FIGURE II.4-1 (Cont'd)

GAIN SEARCH PROCEEDS ALONG A NEW GRADIENT. ITERATION NUMBER 2			
EIGVEC ERROR MESSAGES SW1=0.1207E-02 ITER= 5 DIF=0.0000E 00			
EIGENVECTORS CORRES TO MRP EIGENVALUE			
ROW REAL PART	ROW IMAG PART	COL REAL PART	COL IMAG PART
0.1000000E 01	0.0000000E 00	0.2210334E 00	0.0000000E 00
0.1029001E 00	0.0000300E 00	-0.1373143E-01	0.0000000E 00
-0.9744528E 00	0.0000300E 00	0.1000000E 01	0.0000000E 00
-0.1416449E-01	0.0000000E 00	0.9880193E 00	0.0000000E 00
-0.1529490E-02	0.0000000E 00	-0.6137960E-01	0.0000000E 00
-0.1885111E-01	0.0000000E 00	0.1733184E 00	0.0000000E 00
-0.1896896E-02	0.0000000E 00	-0.1076721E-01	0.0000000E 00
GRADIENT MATRIX -0.5431192E-03 0.3374061E-04			
STABILITY INCREASE STEP SIZE = 0.24275460			
GAIN MATRIX K 0.4054773E 03 -0.4449835E-02			
ROOTS REAL PART IMAG PART			
-0.586954E 01	0.679102E 01		
-0.586954E 01	-0.679102E 01		
0.137404E 01	0.644784E 01		
0.137404E 01	-0.644784E 01		
-0.313356E 01	0.000000E 00		
0.199437E 01	0.000000E 00		
-0.442964E-01	0.000000E 00		
CONVEX CURVE FIT TO STABILITY LOCUS			
GAIN MATRIX K -0.1626357E 02 -0.1829822E 02			
ROOTS REAL PART IMAG PART			
-0.483643E 01	0.549917E 01		
-0.483643E 01	-0.549917E 01		
-0.307465E-01	0.612206E 01		
-0.307465E-01	-0.612206E 01		
-0.157808E 00	0.395643E 00		
-0.157808E 00	-0.395643E 00		
-0.124568E 00	0.000000E 00		

FIGURE II.4-1 (Cont'd)

```

STEP SIZE TOO LARGE. PROGRAM CONTINUES WITH REDUCED STEP SIZE
CONVEX CURVE FIT TO STABILITY LOCUS
GAIN MATRIX K1
-0.2936047E-02  -0.1748459E-02
ROOTS REAL PART IMAG PART
-0.480006E-01  0.542807E-01
-0.480006E-01  -0.542807E-01
-0.107239E-00  0.612655E-01
-0.107239E-00  -0.612655E-01
-0.144420E-00  0.728661E-00
-0.144420E-00  -0.728661E-00
-0.707709E-01  0.000000E-00
MAXIMUM REAL PART OF ROOTS
-0.707709E-01

```

FIGURE II.4-1 (Cont'd)

GAIN SEARCH PROCEEDS ALONG A NEW GRADIENT. ITERATION NUMBER 3			
EIGVEC ERROR MESSAGES			
SNI=0.2492E-02		ITER= 4	DIF=0.2384E-06
EIGENVECTORS CORRES TO MRP EIGENVALUE			
ROW REAL PART	ROW IMAG PART	COL REAL PART	COL IMAG PART
-0.1000000E-01	0.0000000E+00	0.3028623E-00	0.0000000E+00
0.1424851E-00	0.0000000E+00	0.2144598E-01	0.0000000E+00
-0.9601080E+00	0.0000000E+00	0.1000000E-01	0.0000000E+00
0.1390047E-01	0.0000000E+00	0.9848484E-00	0.0000000E+00
-0.2110891E-02	0.0000000E+00	-0.6974274E-01	0.0000000E+00
-0.2838436E-01	0.0000000E+00	0.1727620E-00	0.0000000E+00
-0.2858681E-02	0.0000000E+00	-0.1223427E-01	0.0000000E+00
GRADIENT MATRIX			
-0.1275508E-02		0.9032596E-04	
STABILITY INCREASE STEP SIZE			
		0.00433192	
GAIN MATRIX K			
-0.2598119E-02		-0.1772389E-02	
ROOTS	REAL PART	IMAG PART	
	-0.480912E-01	0.544784E-01	
	-0.480912E-01	-0.544784E-01	
	-0.870506E-01	0.612440E-01	
	-0.870506E-01	-0.612440E-01	
	-0.153032E+00	0.658613E-00	
	-0.153032E+00	-0.658613E-00	
	-0.759749E-01	0.000000E+00	
STEP SIZE IS DOUBLED			
GAIN MATRIX K			
-0.2260190E-02		-0.1796320E-02	
ROOTS	REAL PART	IMAG PART	
	-0.481842E-01	0.546706E-01	
	-0.481842E-01	-0.546706E-01	
	-0.667294E-01	0.612201E-01	
	-0.667294E-01	-0.612201E-01	
	-0.159904E+00	0.580474E-00	
	-0.159904E+00	-0.580474E-00	
	-0.841529E-01	0.000000E+00	

NOT REPRODUCIBLE

FIGURE II.4-1 (Cont'd)

STEP SIZE TOO LARGE. RETURN TO PREVIOUS BEST GAINS		
GAIN MATRIX K		
-0.2598119E+02		
ROOTS	REAL PART	IMAG. PART
-0.480912E-01	-0.544784E-01	
-0.480912E-01	-0.544784E-01	
-0.870506E-01	-0.612440E-01	
-0.870506E-01	-0.612440E-01	
-0.153032E+00	-0.658613E+00	
-0.153032E+00	-0.658613E+00	
-0.759749E-01	-0.000000E+00	

FIGURE II.4-1 (Cont'd)

STABILITY IMPROVEMENT RATE TOO SLOW. PROGRAM TERMINATED.
LAST RESULTS SHOW MOST STABLE CONDITION FOUND.
COMPILE TIME = 2.57 SEC. EXECUTION TIME = 41.39 SEC. UNRECOGNIZED CODE = 40612. Y1

FIGURE EE-4-1 (Cont'd)

In order to verify that GRADGN did indeed attain a stability maximum, the eigenvalues of the system described by (II.4-1) to (II.4-4) were obtained for a matrix of feedback gains. The values of the real parts of the least stable eigenvalues were cross-plotted against the feedback gains, resulting in the equi-stability contours shown in Figure II.4-2. The gain trajectory produced by GRADGN is superimposed on Figure II.4-2. The steepest descent nature of GRADGN, and the fact that a stability maximum was achieved (within the limits of the diminishing returns termination) are evident from Figure II.4-2.

II.5 Comments on Gain Initialization

The digital computer program GRADGN clearly provides a way to improve the stability of a system by output feedback. As such it is a useful adjunct to the output feedback optimization techniques of Cassidy and Levine. It must be remembered, however, that GRADGN is designed to find purely local stability maxima. Thus it may on occasion fail to find feedback gains giving sufficient stability for initialization of the optimization algorithms, even though such gains actually exist. In an attempt to ameliorate this problem, provision has been included in GRADGN for initializing its gain search at any desired location. Thus any desired volume of gain space may be searched by initializing GRADGN at a suitable number of discrete points.

Since GRADGN is designed to perform a similar function to Koenigsberg's²¹ algorithm, a comparison of the two technique's may be pertinent. Both are gradient techniques and both use adaptive step size variation, although Koenigsberg's adaptation procedure is much simpler than that used in GRADGN. A major difference occurs in the computation of the gradient. Koenigsberg's

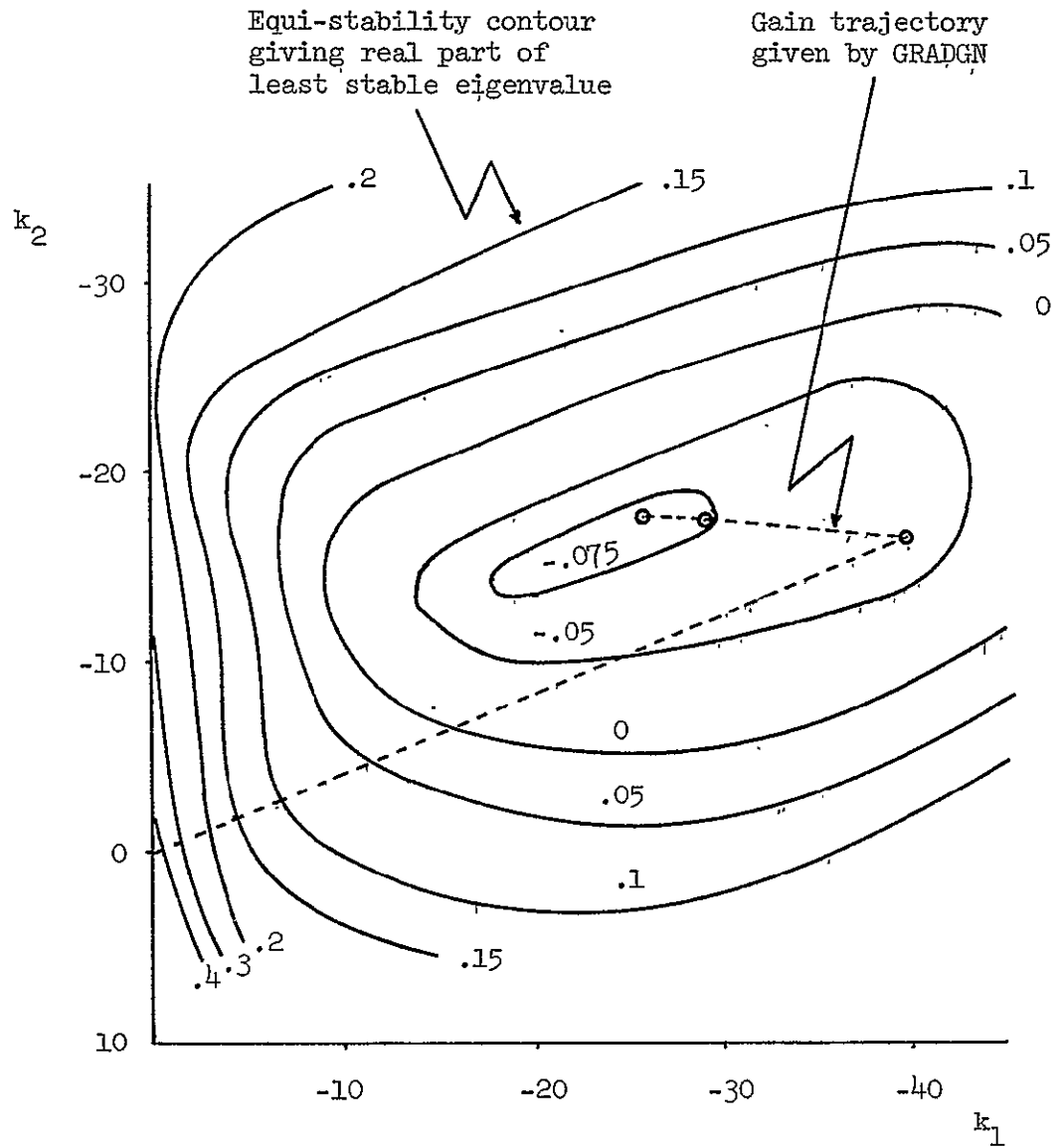


FIGURE II.4-2 Demonstration of GRADGN's
Performance on Example Problem

gradient computation is based on Reddy's³⁰ expression

$$\frac{\partial \lambda_k}{\partial \alpha_p} = \frac{\text{Tr} \left(\text{conjoint} [A] \left[\frac{\partial A}{\partial \alpha_p} \right] \right)}{\text{Tr} \left(\text{conjoint} [A] \right)} \bigg|_{\lambda = \lambda_k} \quad (\text{II.5-1})$$

for the sensitivity of the eigenvalue λ_k of the matrix A to the matrix parameter α_p . The conjoint is defined by

$$\text{conjoint} [A] = \text{adjoint} [A - \lambda I] \quad (\text{II.5-2})$$

The equivalent expression used by GRADGN is (from (II.2-5))

$$\frac{\partial \lambda_k}{\partial \alpha_p} = \frac{\underline{v}^T \frac{\partial A}{\partial \alpha_p} \underline{w}}{\underline{v}^T \underline{w}} \quad (\text{II.5-3})$$

where \underline{v} and \underline{w} respectively are the row and column eigenvectors of the matrix A corresponding to the eigenvalue λ_k . Equation (II.5-3) may be expanded to

$$\frac{\partial \lambda_k}{\partial \alpha_p} = \frac{\sum_{i=1}^{NS} \sum_{j=1}^{NS} v_i \left[\frac{\partial A}{\partial \alpha_p} \right]_{ij} w_j}{\sum_{i=1}^{NS} \sum_{j=1}^{NS} v_i w_j} \quad (\text{II.5-4})$$

which may be reordered to

$$\frac{\partial \lambda_k}{\partial \alpha_p} = \frac{\sum_{j=1}^{NS} \sum_{i=1}^{NS} w_j v_i \left[\frac{\partial A}{\partial \alpha_p} \right]_{ij}}{\sum_{j=1}^{NS} \sum_{i=1}^{NS} w_j v_i} \quad (\text{II.5-5})$$

Inspection of (II.5-5) shows it to be the same as

$$\frac{\partial \lambda_k}{\partial \alpha_p} = \frac{\text{tr} \left[\underline{w} \underline{v}^T \frac{\partial A}{\partial \alpha_p} \right]}{\text{tr} \left[\underline{w} \underline{v}^T \right]} \quad (\text{II.5-6})$$

Now the determination of a column eigenvector of a matrix may be achieved by the computation of only one column of the conjoint. Van Ness,²⁹ EIGVEC, used for eigenvector computation in GRADGN, gives both row and column eigenvectors with about the same computational effort as is required for only a column eigenvector. Thus (II.5-6) is computationally more efficient than (II.5-1). However (II.5-6) is simply (II.5-3) with each vector inner product replaced by the trace of a vector outer product, so that (II.5-3) is clearly preferable to (II.5-6) especially for systems of high order.

CHAPTER III

THE FINITE TIME INTERVAL PROBLEM

III.1 Introduction

In this Chapter we are concerned with the determination of the optimal time invariant output feedback controller for the system

$$\dot{\underline{x}}(t) = A \underline{x}(t) + B \underline{u}(t) \quad (\text{III.1-1})$$

where $\underline{x}(t)$ is a NS vector

$\underline{u}(t)$ is a NC vector

with the cost functional

$$J = \underline{x}^T(T) F \underline{x}(T) + \int_0^T \underline{x}^T(\tau) Q \underline{x}(\tau) + \underline{u}^T(\tau) R \underline{u}(\tau) d\tau \quad (\text{III.1-2})$$

where T is a fixed finite time and F, Q, R are suitably positive (semi) definite. Thus we require

$$\underline{u}(t) = -K^T \underline{x}(t). \quad (\text{III.1-3})$$

and constrain K to be of the form

$$K = \left\{ \begin{array}{c} \text{NC} \\ \overbrace{\left[\begin{array}{c} k_{ij} \\ \vdots \\ 0 \end{array} \right]}^{\text{NF}} \end{array} \right\} \text{NS} \quad (\text{III.1-4})$$

Two items set this problem apart from the normal linear-quadratic state regulator problem. These are the requirement for output feedback control and the requirement for the time-invariant feedback gains. It is remarkable that each of these requirements has been individually satisfied by different investigators, but by the use of similar techniques. Levine¹⁹

used the Initial State Averaging (I.S.A.) technique to determine the optimal time-variable output feedback controller for the system (III.1-1) with the cost functional (III.1-2). On the other hand Kleinman, Fortmann and Athans²⁵ used I.S.A. to determine the optimal time-invariant allstate feedback controller. This report considers the use of I.S.A. to solve both requirements simultaneously.

The theory underlying the I.S.A. approach is presented in Section III.2

III.2 Theory of the Initial State Averaging Technique

The solution to the problem posed by (III.1-1) to (III.1-3) is underdefined in the sense that the optimal controller is a function of the undefined initial condition $\underline{x}(0)$. In general each different initial condition requires a different set of feedback gains. Since no single controller can minimize the cost functional for all initial conditions it seems wise to seek that controller which minimizes the expected value of the cost functional. This is the rationale behind I.S.A.

From (III.1-1) and (III.1-3) we see that

$$\dot{\underline{x}}(t) = \hat{A} \underline{x}(t) \quad (\text{III.2-1})$$

where

$$\hat{A} = \begin{bmatrix} A - BK^T \end{bmatrix} \quad (\text{III.2-1})$$

Thus for a given initial condition $\underline{x}(0)$

$$\underline{x}(t) = \phi(t) \underline{x}(0) \quad (\text{III.2-3})$$

where $\phi(t)$ is the state transition matrix of \hat{A} and is defined by

$$\dot{\phi}(t) = \hat{A} \phi(t) \quad (\text{III.2-4})$$

$$\phi(0) = \mathbf{I} \quad (\text{III.2-5})$$

Since we are dealing only with linear time-invariant systems we have¹

$$\phi(t) = \exp [\hat{\mathbf{A}}t] \quad (\text{III.2-6})$$

Repeating (III.1-2)

$$J = \underline{x}^T(T) \mathbf{F} \underline{x}(T) + \int_0^T \underline{x}^T(\tau) \mathbf{Q} \underline{x}(\tau) + \underline{u}^T(\tau) \mathbf{R} \underline{u}(\tau) d\tau \quad (\text{III.2-7})$$

Using (III.1-3) reduces this to

$$J = \underline{x}^T(T) \mathbf{F} \underline{x}(T) + \int_0^T \underline{x}^T(\tau) [\mathbf{Q} + \mathbf{K}\mathbf{R}\mathbf{K}^T] \underline{x}(\tau) d\tau \quad (\text{III.2-8})$$

whence, by (III.2-3)

$$J = \underline{x}(0)^T \phi^T(T) \mathbf{F} \phi(T) \underline{x}(0) + \int_0^T \underline{x}^T(0) \phi^T(\tau) [\mathbf{Q} + \mathbf{K}\mathbf{R}\mathbf{K}^T] \phi(\tau) \underline{x}(0) d\tau \quad (\text{III.2-9})$$

Moving the time invariant $\underline{x}(0)$ outside the integration reduces (III.2-9)

to

$$J = \underline{x}(0)^T \left[\phi^T(T) \mathbf{F} \phi(T) + \int_0^T \phi^T(\tau) [\mathbf{Q} + \mathbf{K}\mathbf{R}\mathbf{K}^T] \phi(\tau) d\tau \right] \underline{x}(0) \quad (\text{III.2-10})$$

which may be rewritten as

$$J = \underline{x}^T(0) \mathbf{P} \underline{x}(0) \quad (\text{III.2-11})$$

where

$$P = \phi^T(T) F \phi(T) + \int_0^T \phi^T(\tau) [Q + K R K^T] \phi(\tau) d\tau \quad (\text{III.2-12})$$

Thus

$$E[J] = E[x^T(0) P x(0)] \quad (\text{III.2-13})$$

i.e.,

$$E[J] = E \left[\sum_{i=1}^{NS} \sum_{j=1}^{NS} x_i(0) P_{ij} x_j(0) \right] \quad (\text{III.2-14})$$

Now consider the $x_i(0)$, $i = 1, \dots, NS$ to be random variable whose distributions are dependent on the distribution of the initial conditions in state space. Then $E[J]$ is given by (III.2-14) as the expected value of a finite sum of random variables. But the expected value of a finite sum of random variables is equal to the sum of the expected values. Thus

$$E[J] = \sum_{i=1}^{NS} \sum_{j=1}^{NS} E[x_i(0) P_{ij} x_j(0)] \quad (\text{III.2-15})$$

Rearranging (III.2-15) gives

$$E[J] = \sum_{i=1}^{NS} \sum_{j=1}^{NS} E[P_{ij} x_i(0) x_j(0)] \quad (\text{III.2-16})$$

It is clear from (III.2-12) that the P_{ij} in (III.2-16) do not depend on the $x_i(0)$ or the $x_j(0)$. Thus (III.2-16) can be rewritten as

$$E[J] = \sum_{i=1}^{NS} \sum_{j=1}^{NS} P_{ij} E[x_j(0) x_i(0)] \quad (\text{III.2-17})$$

Suppose now that each of the $x_i(0)$ has zero mean and that we can estimate the covariance matrix V of the initial conditions in state space, i.e.,

$$V_{ij} = E[x_i(0) x_j(0)] \quad (\text{III.2-18})$$

Then, from (III.2-17) and (III.2-18)

$$E[J] = \sum_{i=1}^{NS} \sum_{j=1}^{NS} P_{ij} V_{ij} \quad (\text{III.2-19})$$

whence

$$E[J] = \text{Tr} [PV] \quad (\text{III.2-20})$$

Now consider that the system described by (III.1-1), (III.1-3) and (III.1-4) is linear. Thus superposition holds and the system response can be scaled up or down with the magnitude of the initial conditions. This implies that the response of the system to an initial condition lying on the surface of the unit hypersphere in state space characterizes the response to any colinear initial condition state vector. Furthermore it is clear from (III.1-2) that if any initial condition is ratioed by an amount r then the cost is ratioed by r^2 . These two facts imply that the costs associated with the set of initial conditions lying on the surface of the unit hypersphere in state space can be used to conveniently characterize the costs for all initial conditions. It is shown below that with little loss of generality attention may be confined to a uniform distribution of the initial conditions on the surface of the unit hypersphere.

Suppose that the covariance matrix V is non-singular with inverse V^{-1} . Suppose also that the square root of V^{-1} exists, and let it be denoted by $V^{-1/2}$. Consider the description of the system (III.2-1) in terms of a new set of state variables $\underline{y}(t)$ defined by

$$\underline{y}(t) = (\text{NS})^{-1/2} \underline{V}^{-1/2} \underline{x}(t) \quad (\text{III.2-21})$$

Then (III.2-1) becomes

$$(\text{NS})^{1/2} \underline{V}^{1/2} \dot{\underline{y}}(t) = \hat{\underline{A}}(\text{NS})^{1/2} \underline{V}^{1/2} \underline{y}(t) \quad (\text{III.2-22})$$

i.e.,

$$\dot{\underline{y}}(t) = \underline{V}^{-1/2} \hat{\underline{A}} \underline{V}^{1/2} \underline{y}(t) \quad (\text{III.2-23})$$

which is similar in form to (III.2-1).

Also we find that

$$\mathbb{E}[\underline{y}(0) \underline{y}^T(0)] = \mathbb{E}[(\text{NS})^{-1/2} \underline{V}^{-1/2} \underline{x}(0) \underline{x}^T(0) (\text{NS})^{-1/2} \underline{V}^{-1/2}] \quad (\text{III.2-24})$$

Thus

$$\mathbb{E}[\underline{y}(0) \underline{y}^T(0)]_{ij} = (\text{NS})^{-1} \mathbb{E} \left[\sum_{p=1}^{\text{NS}} \sum_{q=1}^{\text{NS}} [\underline{V}^{-1/2}]_{ip} x_p(0) x_q(0) [\underline{V}^{-1/2}]_{qj} \right] \quad (\text{III.2-25})$$

Interchanging finite sum and expectation gives

$$\mathbb{E}[\underline{y}(0) \underline{y}^T(0)]_{ij} = (\text{NS})^{-1} \sum_{p=1}^{\text{NS}} \sum_{q=1}^{\text{NS}} \mathbb{E} \left[[\underline{V}^{-1/2}]_{ip} x_p(0) x_q(0) [\underline{V}^{-1/2}]_{qj} \right] \quad (\text{III.2-26})$$

$$= (\text{NS})^{-1} \sum_{p=1}^{\text{NS}} \sum_{q=1}^{\text{NS}} [\underline{V}^{-1/2}]_{ip} v_{pq} [\underline{V}^{-1/2}]_{qj} \quad (\text{III.2-27})$$

$$= \frac{1}{NS} \left[\begin{array}{ccc} V^{-1/2} & V & V^{-1/2} \end{array} \right]_{ij} \quad (\text{III.2-28})$$

$$= \frac{1}{NS} \left[\begin{array}{c} I \end{array} \right]_{ij} \quad (\text{III.2-29})$$

where I is the identity matrix of order NS .

Thus the covariance matrix of the initial state vector $\underline{y}(0)$ is compatible with a uniform distribution of the $\underline{y}(0)$ vector on the surface of the unit hypersphere. Since such a distribution is convenient and since we can, within the limitations of the non-stringent assumptions made above, transform our initial state vector so that it is compatible with this distribution, we shall hereafter assume that

$$V = \frac{1}{NS} I \quad (\text{III.2-30})$$

Thus, from (III.2-20) we see that

$$E \left[\begin{array}{c} J \end{array} \right] = \frac{1}{NS} \text{tr} \left[\begin{array}{c} P \end{array} \right] \quad (\text{III.2-31})$$

and our I.S.A. optimal control is that which minimizes $\frac{1}{NS} \text{tr} \left[\begin{array}{c} P \end{array} \right]$ where P is given by (III.2-12).

III.3 The Initial State Averaging Algorithm

The objective of the I.S.A. algorithm is to determine that gain matrix K which minimizes the average cost of control as given by (III.2-31) under the constraint that K must be of the form given by (III.1-4).

Thus we wish to minimize the gain function GF given by

$$GF = \frac{1}{NS} \text{tr} \left[\phi^T(T)F \phi(T) + \int_0^T \phi^T(\tau) \left[Q + KRK^T \right] \phi(\tau) d\tau \right] \quad (\text{III.3-1})$$

Now for any two matrices A and B of suitable dimensions we have the trace identity

$$\text{tr} \begin{bmatrix} AB \end{bmatrix} = \text{tr} \begin{bmatrix} BA \end{bmatrix} \quad (\text{III.3-2})$$

Using (III.3-2) and the fact that the integral of a matrix is the matrix of the integrals of the elements reduces (III.3-1) to

$$GF = \frac{1}{NS} \text{tr} \left[F \phi(T) \phi^T(T) + (Q + KRK^T) \int_0^T \phi(\tau) \phi^T(\tau) d\tau \right] \quad (\text{III.3-3})$$

Since the variable elements of the gain matrix K are unconstrained the necessary conditions for K to minimize GF are given by

$$\frac{\partial GF}{\partial k_{ij}} = 0 \quad (\text{III.3-4})$$

$$i = 1, \dots, NF$$

$$j = 1, \dots, NC$$

Expanding (III.3-4) gives the necessary conditions as

$$\begin{aligned} \text{tr} \left[F \left[\frac{\partial \phi}{\partial k_{ij}} \phi^T + \phi \frac{\partial \phi^T}{\partial k_{ij}} \right]_T + \left[\frac{\partial K}{\partial k_{ij}} R K^T + KR \frac{\partial K^T}{\partial k_{ij}} \right] \int_0^T \phi \phi^T d\tau \right. \\ \left. + [Q + KRK^T] \int_0^T \frac{\partial \phi}{\partial k_{ij}} \phi^T + \phi \frac{\partial \phi^T}{\partial k_{ij}} d\tau \right] = 0 \end{aligned} \quad (\text{III.3-5})$$

where the arguments of the ϕ functions have been dropped for clarity.

Now for any matrices A and B

$$\text{tr} \begin{bmatrix} A \end{bmatrix} = \text{tr} \begin{bmatrix} A^T \end{bmatrix} \quad (\text{III.3-6})$$

and

$$\text{tr} [A + B] = \text{tr} [A] + \text{tr} [B] \quad (\text{III.3-7})$$

Thus in (III.3-5)

$$\begin{aligned} \text{tr} \left[F \left[\frac{\partial \phi}{\partial k_{ij}} \phi^T + \phi \frac{\partial \phi^T}{\partial k_{ij}} \right] \right] &= \text{tr} \left[F \frac{\partial \phi}{\partial k_{ij}} \phi^T \right] + \text{tr} \left[F \phi \frac{\partial \phi^T}{\partial k_{ij}} \right] \\ &\quad \text{by (III.3-7)} \\ &= \text{tr} \left[F \frac{\partial \phi}{\partial k_{ij}} \phi^T \right] + \text{tr} \left[\frac{\partial \phi}{\partial k_{ij}} \phi^T F \right] \\ &\quad \text{by (III.3-6)} \\ &= \text{tr} \left[F \frac{\partial \phi}{\partial k_{ij}} \phi^T \right] + \text{tr} \left[F \frac{\partial \phi}{\partial k_{ij}} \phi^T \right] \\ &\quad \text{by (III.3-2)} \\ &= 2 \text{tr} \left[F \frac{\partial \phi}{\partial k_{ij}} \phi^T \right] \quad (\text{III.3-8}) \end{aligned}$$

Similarly

$$\text{tr} \left[\left[\frac{\partial K}{\partial k_{ij}} RK^T + KR \frac{\partial K^T}{\partial k_{ij}} \right] \int_0^T \phi \phi^T d\tau \right] = 2 \text{tr} \left[\frac{\partial K}{\partial k_{ij}} RK^T \int_0^T \phi \phi^T d\tau \right] \quad (\text{III.3-9})$$

and

$$\text{tr} \left[\left[Q + K R K^T \right] \int_0^T \frac{\partial \phi}{\partial k_{ij}} \phi^T + \phi \frac{\partial \phi^T}{\partial k_{ij}} d\tau \right] = 2 \text{tr} \left[\left[Q + K R K^T \right] \int_0^T \phi \frac{\partial \phi^T}{\partial k_{ij}} d\tau \right] \quad (\text{III.3-10})$$

Thus the necessary conditions of (III.3-5) reduce to

$$W = 0 \quad (\text{III.3-11})$$

where the $NF \times NC$ matrix W has its elements defined by

$$w_{ij} = \text{tr} \left[F \phi \frac{\partial \phi^T}{\partial k_{ij}} \right]_T + \left[Q + K R K^T \right] \int_0^T \phi \frac{\partial \phi^T}{\partial k_{ij}} d\tau + \frac{\partial K}{\partial k_{ij}} R K^T \int_0^T \phi \phi^T d\tau \quad (\text{III.3-12})$$

Equations (III.3-11) and (III.3-12) represent a set of $NF \times NC$ simultaneous equations in the $NF \times NC$ variable elements of the gain matrix K . Solution of these equations gives candidates for the optimal controller. The solution is performed by Newton-Raphson iteration.

Consider the rearrangement of W into an $NF \times NC$ vector ${}^n W$ by column ordering

$$\text{i.e.,} \quad w_{ij} = {}^n W_{NF(j-1)+i} \quad (\text{III.3-13})$$

The variable elements of K may be similarly ordered, giving the gain vector ${}^n K$ where

$$K_{ij} = {}^n K_{NF(j-1)+i} \quad (\text{III.3-14})$$

Suppose that ${}^n K(n)$ is the gain vector at the n^{th} iteration of the Newton-Raphson procedure, and that ${}^n W(n)$ is the corresponding

vector of necessary conditions. The Newton-Raphson technique computes $\mathbf{a}_{K(n+1)}$, an improved estimate of a gain vector satisfying the necessary conditions, by

$$\mathbf{a}_{K(n+1)} = \mathbf{a}_{K(n)} - \mu [\nabla(n)]^{-1} \mathbf{a}_{W(n)} \quad (\text{III.3-15})$$

where $\nabla(n)$ is the $(NF \times NC) \times (NF \times NC)$ gradient matrix given by

$$\nabla(n)_{ij} = \frac{\partial \mathbf{a}_{W(n)}^i}{\partial \mathbf{a}_{K(n)}^j} \quad (\text{III.3-16})$$

and μ is a convergence factor.

It can be seen from (III.3-13) and (III.3-14) that computation of $\nabla(n)$ is equivalent to computation of $\frac{\partial \mathbf{a}_{W(n)}^i}{\partial \mathbf{a}_{K(n)}^j}$ for $g = 1, \dots, NF$, $h = 1, \dots, NC$; $p = 1, \dots, NF$; $q = 1, \dots, NC$. From (III.3-12) we have

$$\begin{aligned} \frac{\partial \mathbf{a}_{W(n)}^i}{\partial \mathbf{a}_{K(n)}^j} = & \text{tr} \left[\mathbf{F} \left[\frac{\partial \phi}{\partial \mathbf{k}_{pq}} \frac{\partial \phi^T}{\partial \mathbf{k}_{gh}} + \phi \frac{\partial^2 \phi^T}{\partial \mathbf{k}_{pq} \partial \mathbf{k}_{gh}} \right]_{\mathbf{T}} \right. \\ & + \left[\mathbf{Q} + \mathbf{K} \mathbf{R} \mathbf{K}^T \right] \int_0^T \frac{\partial \phi}{\partial \mathbf{k}_{pq}} \frac{\partial \phi^T}{\partial \mathbf{k}_{gh}} + \phi \frac{\partial^2 \phi^T}{\partial \mathbf{k}_{pq} \partial \mathbf{k}_{gh}} d\tau \\ & + \left[\frac{\partial \mathbf{K}}{\partial \mathbf{k}_{pq}} \mathbf{R} \mathbf{K}^T + \mathbf{K} \mathbf{R} \frac{\partial \mathbf{K}^T}{\partial \mathbf{k}_{pq}} \right] \int_0^T \phi \frac{\partial \phi^T}{\partial \mathbf{k}_{gh}} d\tau + \frac{\partial^2 \mathbf{K}}{\partial \mathbf{k}_{pq} \partial \mathbf{k}_{gh}} \mathbf{R} \mathbf{K}^T \int_0^T \phi \phi^T d\tau \\ & \left. + \frac{\partial \mathbf{K}}{\partial \mathbf{k}_{gh}} \mathbf{R} \left[\frac{\partial \mathbf{K}^T}{\partial \mathbf{k}_{pq}} \int_0^T \phi \phi^T d\tau + \mathbf{K}^T \int_0^T \frac{\partial \phi}{\partial \mathbf{k}_{pq}} \phi^T + \phi \frac{\partial \phi^T}{\partial \mathbf{k}_{pq}} d\tau \right] \right] \end{aligned} \quad (\text{III.3-17})$$

Clearly

$$\frac{\partial^2 K}{\partial k_{pq} \partial k_{gh}} = 0 \quad (\text{III.3-18})$$

Also, by (III.3-6) and (III.3-2)

$$\begin{aligned} \text{tr} \left[KR \frac{\partial K^T}{\partial k_{pq}} \int_0^T \phi \frac{\partial \phi^T}{\partial k_{gh}} d\tau \right] &= \text{tr} \left[\int_0^T \frac{\partial \phi}{\partial k_{gh}} \phi^T d\tau \frac{\partial K}{\partial k_{pq}} RK^T \right] \\ &= \text{tr} \left[\frac{\partial K}{\partial k_{pq}} RK^T \int_0^T \frac{\partial \phi}{\partial k_{gh}} \phi^T d\tau \right] \quad (\text{III.3-19}) \end{aligned}$$

so that (III.3-17) reduces to

$$\begin{aligned} \frac{\partial W_{gh}}{\partial k_{pq}} &= \text{tr} \left[F \left[\frac{\partial \phi}{\partial k_{pq}} \frac{\partial \phi^T}{\partial k_{gh}} + \phi \frac{\partial^2 \phi^T}{\partial k_{pq} \partial k_{gh}} \right]_T \right. \\ &\quad + \left[Q + KRK^T \right] \int_0^T \frac{\partial \phi}{\partial k_{pq}} \frac{\partial \phi^T}{\partial k_{gh}} + \phi \frac{\partial^2 \phi^T}{\partial k_{pq} \partial k_{gh}} d\tau \\ &\quad + \frac{\partial K}{\partial k_{pq}} RK^T \int_0^T \frac{\partial \phi}{\partial k_{gh}} \phi^T + \phi \frac{\partial \phi^T}{\partial k_{gh}} d\tau \\ &\quad \left. + \frac{\partial K}{\partial k_{gh}} RK^T \int_0^T \frac{\partial \phi}{\partial k_{pq}} \phi^T + \phi \frac{\partial \phi^T}{\partial k_{pq}} d\tau + \frac{\partial K}{\partial k_{gh}} R \frac{\partial K^T}{\partial k_{pq}} \int_0^T \phi \phi^T d\tau \right] \quad (\text{III.3-20}) \end{aligned}$$

From (III.3-3), (III.3-12) and (III.3-20) it can be seen that computation of the cost GF, the necessary conditions vector $^a W$ and the gradient matrix ∇ for a given matrix K is dependent on the computation

of a number of functions of the state transition matrix $\phi(t)$. Basic to evaluation of these functions is evaluation of $\frac{\partial \phi(t)}{\partial k_{pq}}$ and $\frac{\partial^2 \phi(t)}{\partial k_{pq} \partial k_{gh}}$ for $p=1, \dots, NF$; $q=1, \dots, NC$; $g=1, \dots, NF$; $h=1, \dots, NC$.

Consider the state transition matrix $\phi_K(t)$ corresponding to a gain matrix \hat{K} given by

$$\hat{K} = K_0 + \frac{\partial K}{\partial k_{pq}} \delta k_{pq} + \frac{\partial K}{\partial k_{gh}} \delta k_{gh} \quad (\text{III.3-21})$$

where δk_{pq} and δk_{gh} are small. By (III.2-4) $\phi_K(t)$ can be computed as the solution to

$$\frac{d \phi_K(t)}{dt} = \left[A - BK_0^T - B \left(\frac{\partial K^T}{\partial k_{pq}} \delta k_{pq} + \frac{\partial K^T}{\partial k_{gh}} \delta k_{gh} \right) \right] \phi_K(t) \quad (\text{III.3-22})$$

with

$$\phi_K(0) = I \quad (\text{III.3-23})$$

Thus

$$\frac{d \phi_K(t)}{dt} = \left[A - BK_0^T \right] \phi_K(t) - B \left[\frac{\partial K^T}{\partial k_{pq}} \delta k_{pq} + \frac{\partial K^T}{\partial k_{gh}} \delta k_{gh} \right] \phi_K(t) \quad (\text{III.3-24})$$

Treating the second term in (III.3-24) as an external input and using (III.2-6)

$$\begin{aligned} \phi_K(t) = & \exp \left[\left[A - BK_0^T \right] t \right] - \int_0^t \exp \left[\left[A - BK_0^T \right] (t-\tau) \right] B \\ & \left[\frac{\partial K^T}{\partial k_{pq}} \delta k_{pq} + \frac{\partial K^T}{\partial k_{gh}} \delta k_{gh} \right] \phi_K(\tau) d\tau \end{aligned} \quad (\text{III.3-25})$$

We now substitute the function described by (III.3-25) for $\phi_K(\tau)$ in (III.3-25). This is similar to the method for computation of the matizant given in DeRusso, Roy and Close¹, and results in

$$\begin{aligned} \phi_K(t) = \phi_{K_0}(t) - \int_0^t \phi_{K_0}(t-\tau) B \left[\frac{\partial K^T}{\partial k_{pq}} \delta k_{pq} + \frac{\partial K^T}{\partial k_{gh}} \delta k_{gh} \right] \\ \left[\phi_{K_0}(\tau) - \int_0^\tau \phi_{K_0}(\tau-s) B \left[\frac{\partial K^T}{\partial k_{pq}} \delta k_{pq} + \frac{\partial K^T}{\partial k_{gh}} \delta k_{gh} \right] \phi_K(s) ds \right] d\tau \end{aligned} \quad (\text{III.3-26})$$

where $\phi_{K_0}(t)$ is the state transition matrix corresponding to the gain matrix K_0 . Thus, to second order in δk_{pq} and δk_{gh} ,

$$\begin{aligned} \phi_K(t) = \phi_{K_0}(t) - \int_0^t \phi_{K_0}(t-\tau) B \left[\frac{\partial K^T}{\partial k_{pq}} \delta k_{pq} + \frac{\partial K^T}{\partial k_{gh}} \delta k_{gh} \right] \\ \left[\phi_{K_0}(\tau) - \int_0^\tau \phi_{K_0}(\tau-s) B \left[\frac{\partial K^T}{\partial k_{pq}} \delta k_{pq} + \frac{\partial K^T}{\partial k_{gh}} \delta k_{gh} \right] \phi_{K_0}(s) ds \right] d\tau \end{aligned} \quad (\text{III.3-27})$$

Expansion of (III.3-27) gives

$$\begin{aligned} \phi_K(t) = \phi_{K_0}(t) - \left[\int_0^t \phi_{K_0}(t-\tau) B \frac{\partial K^T}{\partial k_{pq}} \phi_{K_0}(\tau) d\tau \right] \delta k_{pq} \\ - \left[\int_0^t \phi_{K_0}(t-\tau) B \frac{\partial K^T}{\partial k_{gh}} \phi_{K_0}(\tau) d\tau \right] \delta k_{gh} \end{aligned}$$

$$\begin{aligned}
& + \left[\int_0^t \phi_{K_0}(t-\tau) B \frac{\partial K^T}{\partial k_{pq}} \int_0^\tau \phi_{K_0}(\tau-s) B \frac{\partial K^T}{\partial k_{pq}} \phi_{K_0}(s) ds d\tau \right] \delta^2_{k_{pq}} \\
& + \left[\int_0^t \phi_{K_0}(t-\tau) B \frac{\partial K^T}{\partial k_{pq}} \int_0^\tau \phi_{K_0}(\tau-s) B \frac{\partial K^T}{\partial k_{gh}} \phi_{K_0}(s) ds d\tau \right] \delta_{k_{pq}} \delta_{k_{gh}} \\
& + \left[\int_0^t \phi_{K_0}(t-\tau) B \frac{\partial K^T}{\partial k_{gh}} \int_0^\tau \phi_{K_0}(\tau-s) B \frac{\partial K^T}{\partial k_{pq}} \phi_{K_0}(s) ds d\tau \right] \delta_{k_{pq}} \delta_{k_{gh}} \\
& + \left[\int_0^t \phi_{K_0}(t-\tau) B \frac{\partial K^T}{\partial k_{gh}} \int_0^\tau \phi_{K_0}(\tau-s) B \frac{\partial K^T}{\partial k_{gh}} \phi_{K_0}(s) ds d\tau \right] \delta^2_{k_{gh}}
\end{aligned}
\tag{III.3-28}$$

Now consider that $\phi(t)$ can be expanded in a Taylor's series about K_0 giving

$$\begin{aligned}
\phi_K(t) &= \phi_{K_0}(t) + \frac{\partial \phi(t)}{\partial k_{pq}} \bigg|_{K_0} \delta_{k_{pq}} + \frac{\partial \phi(t)}{\partial k_{gh}} \bigg|_{K_0} \delta_{k_{gh}} \\
&+ \frac{1}{2!} \left[\frac{\partial^2 \phi(t)}{\partial k_{pq}^2} \bigg|_{K_0} \delta^2_{k_{pq}} + 2 \frac{\partial^2 \phi(t)}{\partial k_{pq} \partial k_{gh}} \bigg|_{K_0} \delta_{k_{pq}} \delta_{k_{gh}} \right. \\
&\left. + \frac{\partial^2 \phi(t)}{\partial k_{gh}^2} \bigg|_{K_0} \delta^2_{k_{gh}} \right] \\
&+ \text{terms of higher order}
\end{aligned}
\tag{III.3-29}$$

Comparison of (III.3-28) and (III.3-29) gives

$$\frac{\partial \phi(t)}{\partial k_{pq}} \Big|_{K_0} = - \int_0^t \phi_{K_0}(t-\tau) B \frac{\partial K^T}{\partial k_{pq}} \phi_{K_0}(\tau) d\tau \quad (\text{III.3-30})$$

and

$$\begin{aligned} \frac{\partial^2 \phi(t)}{\partial k_{pq} \partial k_{gh}} \Big|_{K_0} &= \int_0^t \phi_{K_0}(t-\tau) B \frac{\partial K^T}{\partial k_{pq}} \int_0^\tau \phi_{K_0}(\tau-s) \frac{\partial K^T}{\partial k_{gh}} \phi_{K_0}(s) ds d\tau \\ &+ \int_0^t \phi_{K_0}(t-\tau) B \frac{\partial K^T}{\partial k_{gh}} \int_0^\tau \phi_{K_0}(\tau-s) \frac{\partial K^T}{\partial k_{pq}} \phi_{K_0}(s) ds d\tau \end{aligned} \quad (\text{III.3-31})$$

The result given by (III.3-30) agrees with a similar expression presented by Levine and Athans³¹ for the semi-infinite time interval problem.

We note from (III.3-30) and (III.3-31) that

$$\frac{\partial^2 \phi(t)}{\partial k_{pq} \partial k_{gh}} \Big|_{K_0} = - \int_0^t \phi_{K_0}(t-\tau) B \left[\frac{\partial K^T}{\partial k_{pq}} \frac{\partial \phi(t)}{\partial k_{gh}} \Big|_{K_0} + \frac{\partial K^T}{\partial k_{gh}} \frac{\partial \phi(t)}{\partial k_{pq}} \Big|_{K_0} \right] d\tau \quad (\text{III.3-32})$$

The functions of the state transition matrix to be evaluated are now seen to be given by

$$\phi \frac{\partial \phi^T}{\partial k_{pq}} \Big|_T = - \phi(T) \int_0^T \phi^T(\tau) \frac{\partial K}{\partial k_{pq}} B^T \phi^T(T-\tau) d\tau \quad (\text{III.3-33})$$

$$\frac{\partial \phi}{\partial k_{pq}} \frac{\partial \phi^T}{\partial k_{gh}} \Big|_T = \int_0^T \phi(T-\tau) B \frac{\partial K^T}{\partial k_{pq}} \phi(\tau) d\tau \int_0^T \phi^T(s) \frac{\partial K}{\partial k_{gh}} B^T \phi^T(T-s) ds \quad (\text{III.3-34})$$

$$\left. \phi \frac{\partial^2 \phi^T}{\partial k_{pq} \partial k_{gh}} \right|_T = -\phi(T) \int_0^T \left[\frac{\partial \phi^T(\tau)}{\partial k_{pq}} \frac{\partial K}{\partial k_{gh}} + \frac{\partial \phi^T(\tau)}{\partial k_{pq}} \frac{\partial K}{\partial k_{pq}} \right] B^T \phi(T-\tau) d\tau \quad (\text{III.3-35})$$

$$\int_0^T \phi \frac{\partial \phi^T}{\partial k_{pq}} d\tau = - \int_0^T \phi(\tau) \int_0^\tau \phi^T(s) \frac{\partial K}{\partial k_{pq}} B^T \phi^T(\tau-s) ds d\tau \quad (\text{III.3-36})$$

$$\int_0^T \frac{\partial \phi}{\partial k_{pq}} \frac{\partial \phi^T}{\partial k_{gh}} d\tau = \int_0^T \int_0^\tau \phi(\tau-\nu) B \frac{\partial K^T}{\partial k_{pq}} \phi(\nu) d\nu \quad (\text{III.3-37})$$

$$\int_0^\tau \phi^T(s) \frac{\partial K}{\partial k_{gh}} B^T \phi^T(\tau-s) ds d\tau$$

and

$$\int_0^T \phi \frac{\partial^2 \phi^T}{\partial k_{pq} \partial k_{gh}} d\tau = - \int_0^T \phi(\tau) \int_0^\tau \left[\frac{\partial \phi(s)}{\partial k_{pq}} \frac{\partial K}{\partial k_{gh}} + \frac{\partial \phi(s)}{\partial k_{gh}} \frac{\partial K}{\partial k_{pq}} \right] B^T \phi(\tau-s) ds d\tau \quad (\text{III.3-38})$$

Consider (III.3-33) as being representative of (III.3-33) through (III.3-38). Supposing \hat{A} to have distinct eigenvalues we can evaluate $\phi(t)$ by

$$\phi(t) = M \exp \left[\mathcal{A} t \right] M^{-1} \quad (\text{III.3-39})$$

where M is the modal matrix of eigenvectors of \hat{A}

\mathcal{A} is the diagonal matrix of eigenvalues of \hat{A}

Substitution of (III.3-39) in (III.3-33) yields

$$\left. \phi \frac{\partial \phi^T}{\partial k_{pq}} \right|_T = - M e^{\mathcal{A} T} M^{-1} \int_0^T \left[M^{-1} \right]^T e^{\mathcal{A} \tau} M^T \frac{\partial K}{\partial k_{pq}} B^T \left[M^{-1} \right]^T e^{(T-\tau)} M^T d\tau \quad (\text{III.3-40})$$

Now $\frac{\partial K}{\partial k_{pq}}$ is simply an NS x NC matrix having a 1.0 in its pq position and zeros elsewhere. Thus the ij element of $\phi \frac{\partial \phi^T}{\partial k_{pq}} \bigg|_T$ is given by

$$\left[\phi \frac{\partial \phi^T}{\partial k_{pq}} \bigg|_T \right]_{ij} = - \sum_{\substack{n_1, n_2, n_3 \\ n_4, n_5=1}}^{NS} M_{in_1} e^{\lambda_{n_1} T} M_{n_1 n_2}^{-1} \int_0^T M_{n_3 n_2}^{-1} e^{\lambda_{n_3} \tau} M_{pn_3} B_{n_4 q} d\tau$$

$$M_{n_5 n_4}^{-1} e^{\lambda_{n_5} (T-\tau)} M_{jn_5} d\tau \quad (III.3-41)$$

where λ_{n_1} is the n_1 th eigenvalue of A.

Performing the integration yields

$$\left[\phi \frac{\partial \phi^T}{\partial k_{pq}} \bigg|_T \right]_{ij} = - \sum_{\substack{n_1, n_2, n_3 \\ n_4, n_5=1}}^{NS} M_{in_1} M_{n_1 n_2}^{-1} M_{n_3 n_2}^{-1} M_{pn_3} B_{n_4 q} M_{n_5 n_4}^{-1} M_{jn_5}$$

$$\left[\begin{array}{l} \text{iff} \\ n_5=n_3 \end{array} T e^{(\lambda_{n_1}+\lambda_{n_5})T} + \begin{array}{l} \text{iff} \\ n_5 \neq n_3 \end{array} \frac{e^{(\lambda_{n_1}+\lambda_{n_3})T} - e^{(\lambda_{n_1}+\lambda_{n_5})T}}{\lambda_{n_3} - \lambda_{n_5}} \right] \quad (III.3-42)$$

Thus (III.3-33) can be evaluated in terms of the eigenvalues and eigenvectors of A. Equations (III.3-34) through (III.3-38) can be similarly evaluated. The terms can then be combined by (III.3-3), (III.3-12) and (III.3-20) to give the average cost GF, the necessary conditions vector W^0 and the gradient matrix ∇ respectively.

The only other item required for application of the Newton-Raphson iteration technique described by (III.3-15) is a value for the convergence

factor μ . Determination of a suitable value of the convergence factor is described in Section III.4.

III.4 Mechanization of the Initial State Averaging Algorithm

The algorithm described above for computation of optimal constant output feedback gains has been mechanized in the digital computer program ISAFIT. The program was written in Fortran IV for an IBM 360/50 computer. A listing is given in Appendix II. The program comprises a MAIN and several subprograms. The functions of the major subprograms are described briefly below.

MAIN: performs most input-output functions and acts as the controller for the other subprograms. The MAIN subprogram also computes a suitable value for the Newton-Raphson convergence factor μ (see (III.3-15) and controls program termination. The operation is illustrated by the data flowchart given in Figure III.4-1.

Computation of the convergence factor μ is performed adaptively. A trial step is made with μ set to 1.0. The value of the gain function GF for the new gains is compared with its former value. If a decrease has occurred the program proceeds to the next iteration. Otherwise the value of μ is halved. This process is repeated up to five times. If no improvement in the gain function has resulted by the fifth cycle, i.e., $\mu = 2^{-5}$, the program is terminated due to poor convergence.

Assuming no termination due to poor convergence, normal program termination will occur when the feedback gains approach their final values. The criterion for this termination is that the proportional change in each individual gain during a single iteration should be less than an amount GNSTOP. The parameter GNSTOP is input by the user.

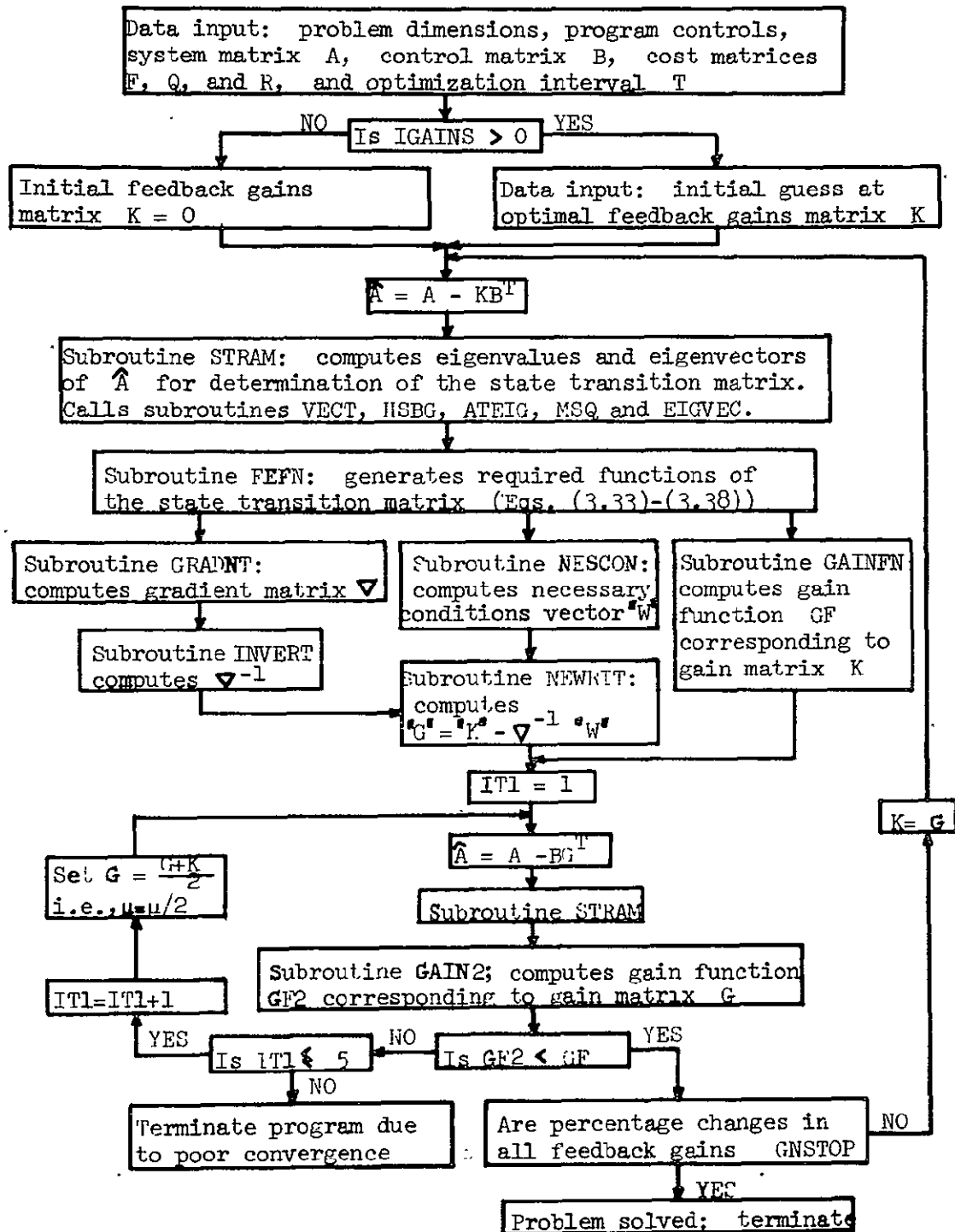


FIGURE III.4-1 Data Flowchart for Digital Computer
Program ISAFT

STRAM: computes the eigenvalues and the model matrix of the matrix \hat{A} . The subprograms VECT, HSBG and ATEIG are used in the eigenvalue determination. The subprograms MSQ and EIGVEC are used to determine the model matrix and its inverse.

FEFN: computes the functions of the state transition matrix described by (III.3-33) through (III.3-38).

GAINFN: computes the value of the gain function GF at the beginning of each iteration - see (III.3-3).

NESCON: computes the necessary conditions vector ${}^nW^n$ - see (III.3-12).

GRADNT: computes the gradient matrix ∇ - see (III.3-20)..

INVERT: inverts the gradient matrix

NEWRTT: performs one iteration of the Newton-Raphson algorithm - see (III.3-15).

GAIN2: computes the value of the gain function GF at the end of each iteration for use in the determination of a suitable convergence factor μ - see (III.3-15).

III.5 Examples of the Use of the Optimization Program ISAFIT

Three examples are given to illustrate the use of the digital computer program ISAFIT. In keeping with the expository nature of this Chapter the examples are all short and are designed to illustrate different aspects of the computations.

Example 1

This example considered the case where all system states were available for feedback. The system equations were

$$\begin{bmatrix} \dot{x}_1(t) \\ \dot{x}_2(t) \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix} \begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u(t) \quad (\text{III.5-1})$$

and the cost functional was

$$J = \underline{x}^T(1) \begin{bmatrix} 2 & 0 \\ 0 & 1 \end{bmatrix} \underline{x}(1) + \int_0^1 \underline{x}^T(\tau) \begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix} \underline{x}(\tau) + 0.1 u^2(\tau) d\tau \quad (\text{III.5-2})$$

The control was constrained to be of the form

$$u(t) = -k_1 x_1(t) - k_2 x_2(t) \quad (\text{III.5-3})$$

It was assumed that there was no prior knowledge about the system, and the initial feedback gains were set to the default value of zero. The computer print out is given in Figure III.5-1. The optimal feedback gains were computed to be

$$k_1 = 2.30 \quad (\text{III.5-4})$$

$$k_2 = 3.42 \quad (\text{III.5-5})$$

To verify that the values given by (III.5-4) and (III.5-5) were indeed optimal the results were checked by independent means. A matrix of feedback gains in the vicinity of the solution gains was chosen. At each set of gains the system described by (III.5-1) and (III.5-3) was simulated over the optimization time interval, starting from a number of different initial conditions, and the costs were computed numerically using (III.5-2). The average cost was then computed for each set of feedback gains. The initial conditions were chosen to be equally spaced around the unit circle in state space. Such an independent check will hereafter be referred to as an I.C. Cross-plotting the I.C. data resulted in the equal-cost contours shown in Figure III.5-2. The gain trajectory produced by ISFT is superimposed


```

ITERATION NUMBER      1

GAIN MATRIX
  0.0000000  0.0000000

SYSTEM EIGENVALUES
  -0.20000000 01  -0.00000000 00
  -0.10000000 01  -0.00000000 00

AVERAGE COST =      0.1227681D 01

NECESSARY CONDITIONS VECTOR
  -0.6159624D-01  -0.9331765D-01

GRADIENT MATRIX
  0.2239452D 00  -0.9323973D-01  -0.9323973D-01  0.1338320D 00

INVERSE GRADIENT MATRIX
  8.2898689      4.3821046      4.3821046      10.5250361

```

FIGURE III.5-1 (Cont'd)

ITERATION NUMBER

2

NEW GAINS

C.7563600

1.2520928

SYSTEM EIGENVALUES

-C.3438649D 01

-0.0000000D 00

-C.8131438D 00

-0.0000000D 00

GAIN TOLERANCE ACHIEVED =

1.0000000

REQUIRED STOPPING TOLERANCE =

0.0100000

AVERAGE COST =

0.1121653D 01

NECESSARY CONDITIONS VECTOR

-C.2673182D-01

-C.3301758D-01

GRADIENT MATRIX

C.1144905D 00

-0.5721144D-01

-0.5721144D-01

0.6673589D-01

INVERSE GRADIENT MATRIX

15.2801933

13.0994265

13.0994265

26.2143374

FIGURE III.5-1 (Cont'd)

```

ITERATION NUMBER          3

NEW GAINS
    1.4372327          2.4677983

SYSTEM EIGENVALUES
    -C.4692691D 01      -C.0C0C0C0D 00
    -C.7751C72D 00      -C.0C0C0C0D 00

    GAIN TOLERANCE ACHIEVED =          0.5136254

    REQUIRED STOPPING TOLERANCE =        0.0100000

    AVERAGE COST *          C.1083638D 01

    NECESSARY CONDITIONS VECTOR
        -C.8372973D-C2      -C.9586460D-02

    GRADIENT MATRIX
        C.6909313D-01      -0.3735754D-01      -0.3735754D-01      0.3910556D-01

    INVERSE GRADIENT MATRIX
        29.9352104          28.5971025          28.5971025          52.8906197

```

FIGURE III.5-1 (Cont'd)

ITERATION NUMBER

NEW CAINS

2.16213C4	3.2142749
-----------	-----------

SYSTEM EIGENVALUES

-C.5450676D C1	-C.0000000D 00
-C.7635589D C0	-C.0000000D 00

GAIN TOLERANCE ACHIEVED =	0.2427197
---------------------------	-----------

REGLIRED STOPPING TOLERANCE =	0.0100000
-------------------------------	-----------

AVERAGE COST =	C.1C77172D 01
----------------	---------------

NECESSARY CONDITIONS VECTOR

-C.1332613D-C2	-C.1736675D-02
----------------	----------------

GRACIENT MATRIX

C.5362835D-C1	-0.2976007D-01	-0.2976007D-01	0.2943540D-01
---------------	----------------	----------------	---------------

INVERSE GRADIENT MATRIX

42.4808458	42.9494143	42.9494143	77.3958539
------------	------------	------------	------------

```

ITERATION NUMBER      5

NEW GAINS
  2.2933385          3.4059298

SYSTEM EIGENVALUES
  -C.5645432D 01    -C.0000000D 00
  -C.7604978D 00    -C.0000000D 00

GAIN TOLERANCE ACHIEVED =      0.0572127

REQUIRED STOPPING TOLERANCE =    0.0100000

AVERAGE COST =      0.1076910D 01

NECESSARY CONDITIONS VECTOR
  -C.4792496D-04    -C.8692727D-04

GRADIENT MATRIX
  C.5059286D-01    -0.2817238D-01    -0.2817238D-01    0.2745821D-01

INVERSE GRADIENT MATRIX
  46.1089873      47.3082380      47.3082380      84.9576547

```

FIGURE III.5-1 (Cont'd)

ITERATION NUMBER 6

NEW GAINS
2.2996607 3.4155822

SYSTEM EIGENVALUES
-C.5655292D C1 -0.0000000D 00
-C.7602897D 00 -0.0000000D 00

GAIN TOLERANCE ACHIEVED = 0.0028260

REQUIRED STOPPING TOLERANCE = 0.0100000

SOLUTION IS COMPLETE. ABOVE GAINS ARE OPTIMAL

AVERAGE COST = 0.1076909D 01

COMPILE TIME= 28.40 SEC, EXECUTION TIME= 193.64 SEC, OBJECT CODE= 64192 BYTES,

FIGURE III.5-1 (Cont'd)

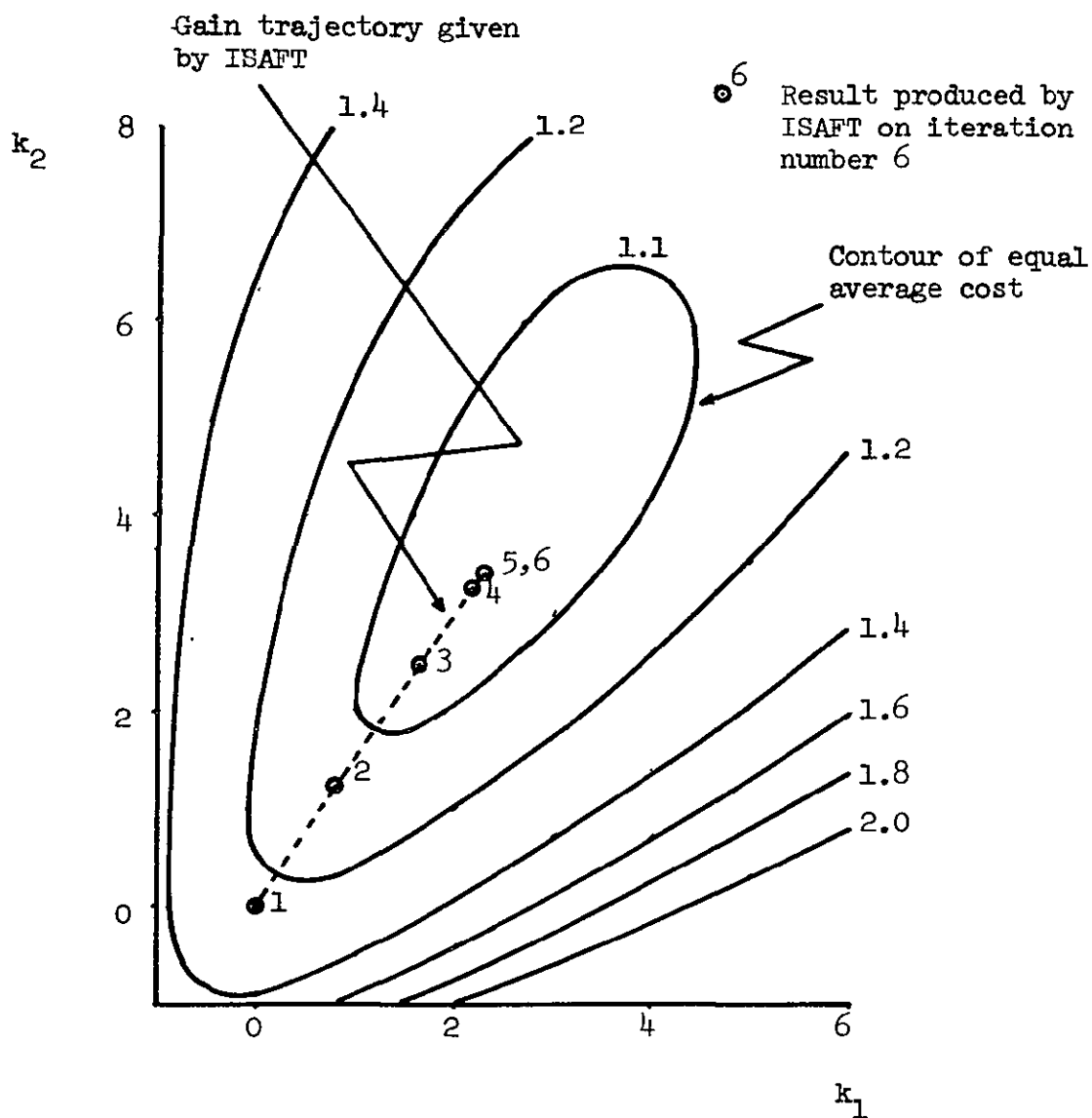


FIGURE III.5-2 Verification of the ISAFIT results for Example 1

on Figure III.5-2. It is apparent that ISAFIT determined feedback gains giving a minimum average cost.

To provide a comparison for the results produced by ISAFIT, the problem defined by (III.5-1) and (III.5-2) was solved for the optimal time varying feedback gains by the normal procedure of Ricatti matrix backwards integration.² The resulting feedback gains are compared in Figure III.5-3, with those produced by ISAFIT. The average cost for the optimal time varying gains was computed by numerical integration from a number of initial conditions. The average cost for the time-varying gains was 1.030 compared with 1.077 for the time invariant gains.

Example 2

This example used the same system equations and cost functional as Example 1 - see (III.5-1) and (III.5-2) respectively. The feedback structure however, was constrained to feedback of only the first state i.e.,

$$u(t) = -k_1 x_1(t) \quad (\text{III.5-6})$$

A rather inaccurate guess at the optimal feedback gain resulted in the computer print out given in Figure III.5-4. The use of the convergence factor is evident.

Figure III.5-5 compares the results given by ISAFIT with those determined by an I.C. It is again clear that ISAFIT determined a feedback gain giving a minimum average cost.

The minimum average cost for the single time-invariant feedback was 1.219 compared with 1.077 for allstate time invariant feedback and 1.030 for allstate time varying feedback.

Example 3

This example was chosen to check the effect of complex eigenvalues

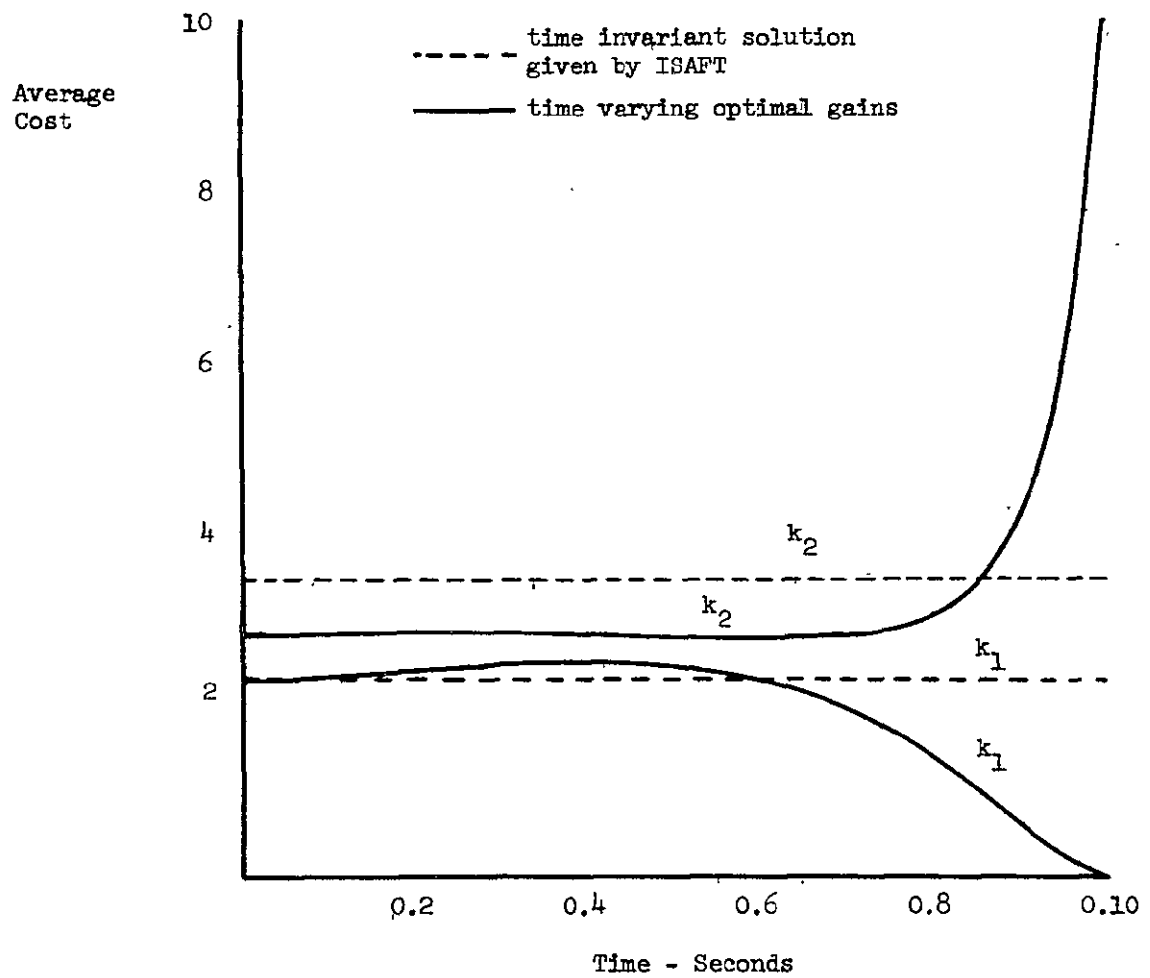


FIGURE III.5-3 Comparison of Time-varying and Time-invariant Optimal Feedback Gains for Example 1

STATES	CONTROLS	FEEDBACKS	GAINS
2	1	1	1

SYSTEM MATRIX A				
0.0000000	1.0000000	-2.0000000	-3.0000000	

CONTROL MATRIX B
0.000000 1.000000

TERMINAL TIME T = 1.0000000

TERMINAL COST MATRIX F			
2.0000000	0.0000000	0.0000000	1.0000000

STATE WEIGHTING MATRIX Q			
1.0000000	0.0000000	0.0000000	2.0000000

CONTROL WEIGHTING MATRIX R:
0.1000000

ITERATION NUMBER

1

GAIN MATRIX

5.0000000

SYSTEM EIGENVALUES

-0.15000000 01 0.21794490 01

-0.15000000 01 -0.21794490 01

AVERAGE COST =

0.21604570 01

NECESSARY CONDITIONS VECTOR

0.27291080 00

GRADIENT MATRIX

0.98928970-02

INVERSE GRADIENT MATRIX

101.0826292

FIGURE III.5-4 (Cont'd)

ITERATION NUMBER

2

NEW GAINS

-22.5865428

SYSTEM EIGENVALUES

-0.62787600 01 -0.00000000 00
 0.32787600 01 0.00000000 00

STEP SIZE TOO LARGE, NEW AVERAGE COST WOULD HAVE BEEN 0.37338830 04
 GAIN ADJUSTMENT IS HALVED

NEW GAINS

-8.7932714

SYSTEM EIGENVALUES

-0.45072030 01 -0.00000000 00
 0.15072030 01 0.00000000 00

STEP SIZE TOO LARGE, NEW AVERAGE COST WOULD HAVE BEEN 0.50750500 02
 GAIN ADJUSTMENT IS HALVED

NEW GAINS

-1.8966357

SYSTEM EIGENVALUES

-0.29651400 01 -0.00000000 00
 -0.34859840 01 -0.00000000 00

GAIN TOLERANCE ACHIEVED = 3.6362469

REQUIRED STOPPING TOLERANCE = 0.0100000

AVERAGE COST = 0.19246870 01

NECESSARY CONDITIONS VECTOR

-0.78732140 00

GRADIENT MATRIX

0.58702700 00

FIGURE III.5-4 (Cont'd)

INVERSE GRADIENT MATRIX
1.7034993



FIGURE III.5-4 (Cont'd)

ITERATION NUMBER

3

NEW GAINS

-0.5554343

SYSTEM EIGENVALUES

-0.2397460D 01 -0.0000000D 00

-0.6025401D 00 -0.0000000D 00

GAIN TOLERANCE ACHIEVED = 2.4146897

REQUIRED STOPPING TOLERANCE = 0.0100000

AVERAGE COST = 0.1300264D 01

NECESSARY CONDITIONS VECTOR

-0.2070736D 00

GRADIENT MATRIX

0.3030369D 00

INVERSE GRADIENT MATRIX

3.2999278

FIGURE III.5-4 (Cont'd)

ITERATION NUMBER

NEW GAINS
0.1278936

SYSTEM EIGENVALUES
-0.1849437D 01 -0.00000000 00
-0.1150563D 01 -0.00000000 00

GAIN TOLERANCE ACHIEVED = 5.3429401

REQUIRED STOPPING TOLERANCE = 0.0100000

AVERAGE COST = 0.1221592D 01

NECESSARY CONDITIONS VECTOR
-0.3396536D-01

GRADIENT MATRIX
0.2082957D 00

INVERSE GRADIENT MATRIX
4.8008683

FIGURE III.5-4 (Cont'd)

```

ITERATION NUMBER      5
NEW GAINS
  0.2909568
SYSTEM EIGENVALUES
-0.15000000 01      0.20237790 00
-0.15000000 01      -0.20237790 00
GAIN TOLERANCE ACHIEVED = 0.5604378
REQUIRED STOPPING TOLERANCE = 0.0100000
AVERAGE COST = 0.12187380 01
NECESSARY CONDITIONS VECTOR
  0.15424180 02
GRADIENT MATRIX
  0.18960870 00
INVERSE GRADIENT MATRIX
  5.2740210

```

FIGURE III.5-4 (Cont'd)

ITERATION NUMBER

6

NEW GAINS

0.2990916

SYSTEM EIGENVALUES

-0.15000000 01 -0.22156610 00

-0.15000000 01 -0.22156610 00

GAIN TOLERANCE ACHIEVED =

0.0271982

REQUIRED STOPPING TOLERANCE =

0.0100000

AVERAGE COST

0-12187320-01

NECESSARY CONDITIONS. VECTOR

-0.36478570-05

GRADIENT MATRIX

0.18871240 .00

INVERSE GRADIENT MATRIX

5.2990702

FIGURE III.5-4 (Cont'd)

ITERATION NUMBER 7

NEW GAINS
0.2991109

SYSTEM EIGENVALUES
-0.15000000 01 0.22160980 00
-0.15000000 01 -0.22160980 00

GAIN TOLERANCE ACHIEVED = 0.0000646

REQUIRED STOPPING TOLERANCE = 0.0100000

SOLUTION IS COMPLETE. ABOVE GAINS ARE OPTIMAL

AVERAGE COST = 0.12187320 01

COMPILE TIME= 27.14 SEC, EXECUTION TIME= 69.23 SEC, OBJECT CODE= 64216 BYTES

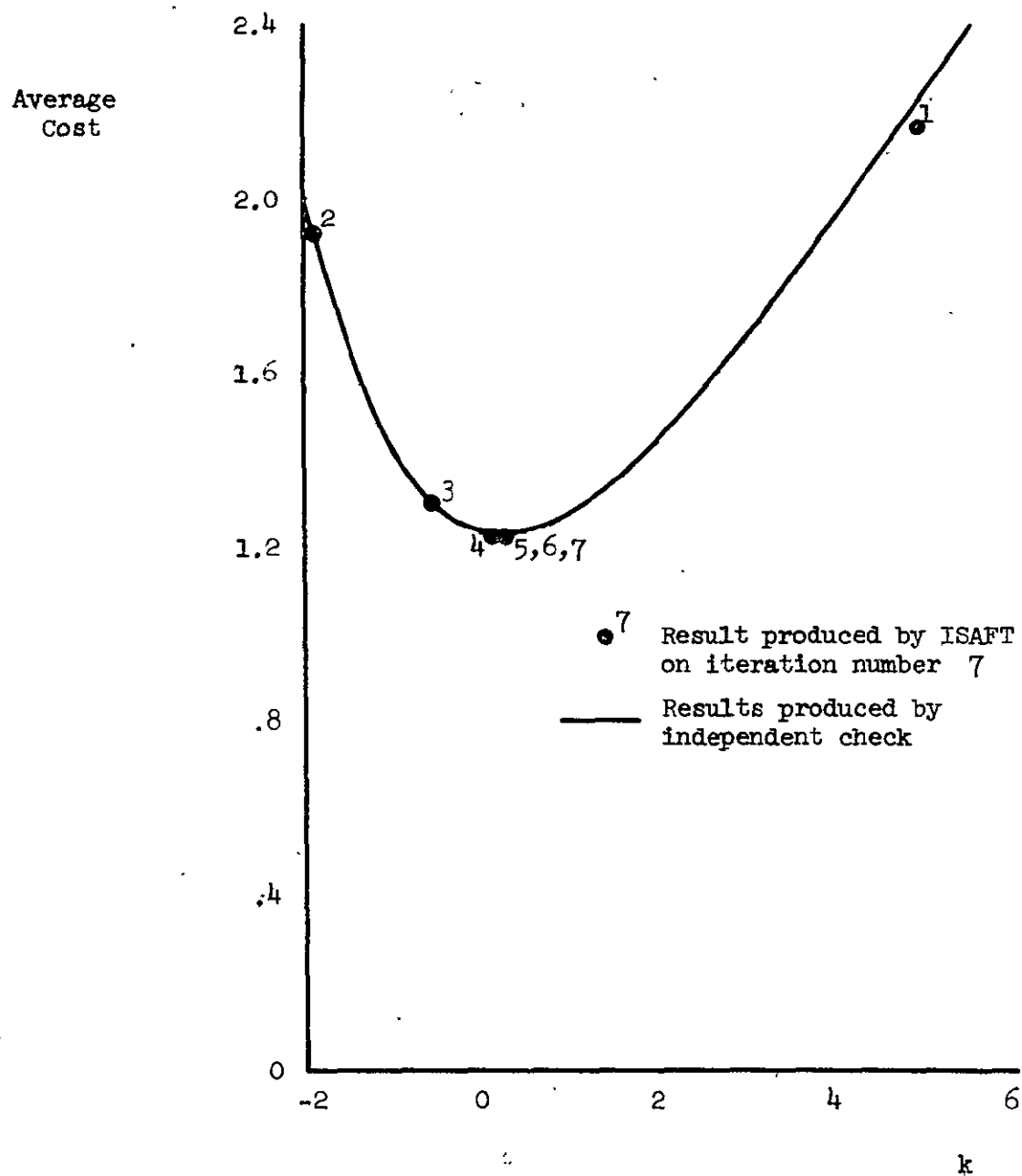


FIGURE III.5-5 Verification of the ISAFIT results
for Example 2

on the computations. The system equations were

$$\begin{bmatrix} \dot{x}_1(t) \\ \dot{x}_2(t) \\ \dot{x}_3(t) \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 6 & 7 & 0 \end{bmatrix} \begin{bmatrix} x_1(t) \\ x_2(t) \\ x_3(t) \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} u(t) \quad (\text{III.5-7})$$

with the cost functional

$$J = \underline{x}^T(1) \begin{bmatrix} 3 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 1 \end{bmatrix} \underline{x}(1) + \int_0^1 \underline{x}^T(\tau) \begin{bmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 3 \end{bmatrix} \underline{x}(\tau) + 0.1 u^2(\tau) d\tau \quad (\text{III.5-8})$$

It was assumed that only the first state was available for feedback so that

$$u(t) = -k_1 x_1(t) \quad (\text{III.5-9})$$

The computer print-out for a starting gain of zero is given in Figure III.5-6 and is compared with an I.C. in Figure III.5-7. There is an offset in Figure III.5-7 between the average costs produced by ISAFIT and those produced by the I.C. This is due to the difficulty in approximating a uniform distribution on the surface of a sphere by a finite number of discrete points. (This difficulty increases rapidly with the order of the system). It is nevertheless clear that ISAFIT achieved a minimum average cost.

III.6 Comments on the Finite Time Problem

The objective of the I.S.A. approach is to determine the feedback gains giving the minimum value of the gain function GF, which is the expected value of the cost functional. The I.S.A. algorithm presented in Section III.3, however, is designed only to determine feedback gains

STATES	CONTROLS	FEEDBACKS	IGAINS
3	1	1	0

SYSTEM MATRIX A			
0.0000000	1.0000000	0.0000000	0.0000000
0.0000000	1.0000000	0.0000000	7.0000000
0.0000000			

CONTROL MATRIX B		
0.0000000	0.0000000	1.0000000

TERMINAL TIME T. = 1.0000000

TERMINAL COST MATRIX F			
3.0000000	0.0000000	0.0000000	0.0000000
2.0000000	0.0000000	0.0000000	0.0000000
1.0000000			

STATE WEIGHTING MATRIX Q			
1.0000000	0.0000000	0.0000000	0.0000000
2.0000000	0.0000000	0.0000000	0.0000000
3.0000000			

CONTROL WEIGHTING MATRIX R
0.1000000

FIGURE III.5F6 Computer Print-Out For Example 3


```

ITERATION NUMBER      1
GAIN MATRIX
  0.0000000
SYSTEM EIGENVALUES
  0.3000000D 01      0.0000000D 00
 -0.2000000D 01      0.0000000D 00
 -0.1000000D 01      0.0000000D 00
AVERAGE COST =      0.6821096D 03
NECESSARY CONDITIONS VECTOR
 -0.1782183D 03
GRADIENT MATRIX
  0.2884249D 02
INVERSE GRADIENT MATRIX
  0.0346711

```

FIGURE III.5-6 (Cont'd)

ITERATION NUMBER	2
NEW GAINS	6.1790200
SYSTEM EIGENVALUES	
-0.2658447D-01	-0.0000000D-00
0.2632870D-01	0.0000000D-00
0.2557668D-01	0.0000000D-00
GAIN TOLERANCE ACHIEVED	1.0000000
REQUIRED STOPPING TOLERANCE	0.0500000
AVERAGE COST	0.2563643D-03
NECESSARY CONDITIONS VECTOR	-0.4158308D-02
GRADIENT MATRIX	0.1611207D-02
INVERSE GRADIENT MATRIX	0.0620653

FIGURE III.5-6 (Cont'd)

ITERATION NUMBER 4

NEW GAINS 9.2114946

SYSTEM EIGENVALUES

-0.28507120 01	-0.00000000 00
0.23767130 01	0.00000000 00
0.47399860 00	0.00000000 00

GAIN TOLERANCE ACHIEVED = 0.0490268

REQUIRED STOPPING TOLERANCE = 0.0500000

SOLUTION IS COMPLETE. ABOVE GAINS ARE OPTIMAL

AVERAGE COST = 0.21660910 03

COMPUTE TIME = 26.74 SEC. EXECUTION TIME = 976.35 SEC. OBJECT CODE = 64216 BYTES

NOT REPRODUCIBLE

FIGURE III.5-6 (Cont'd)

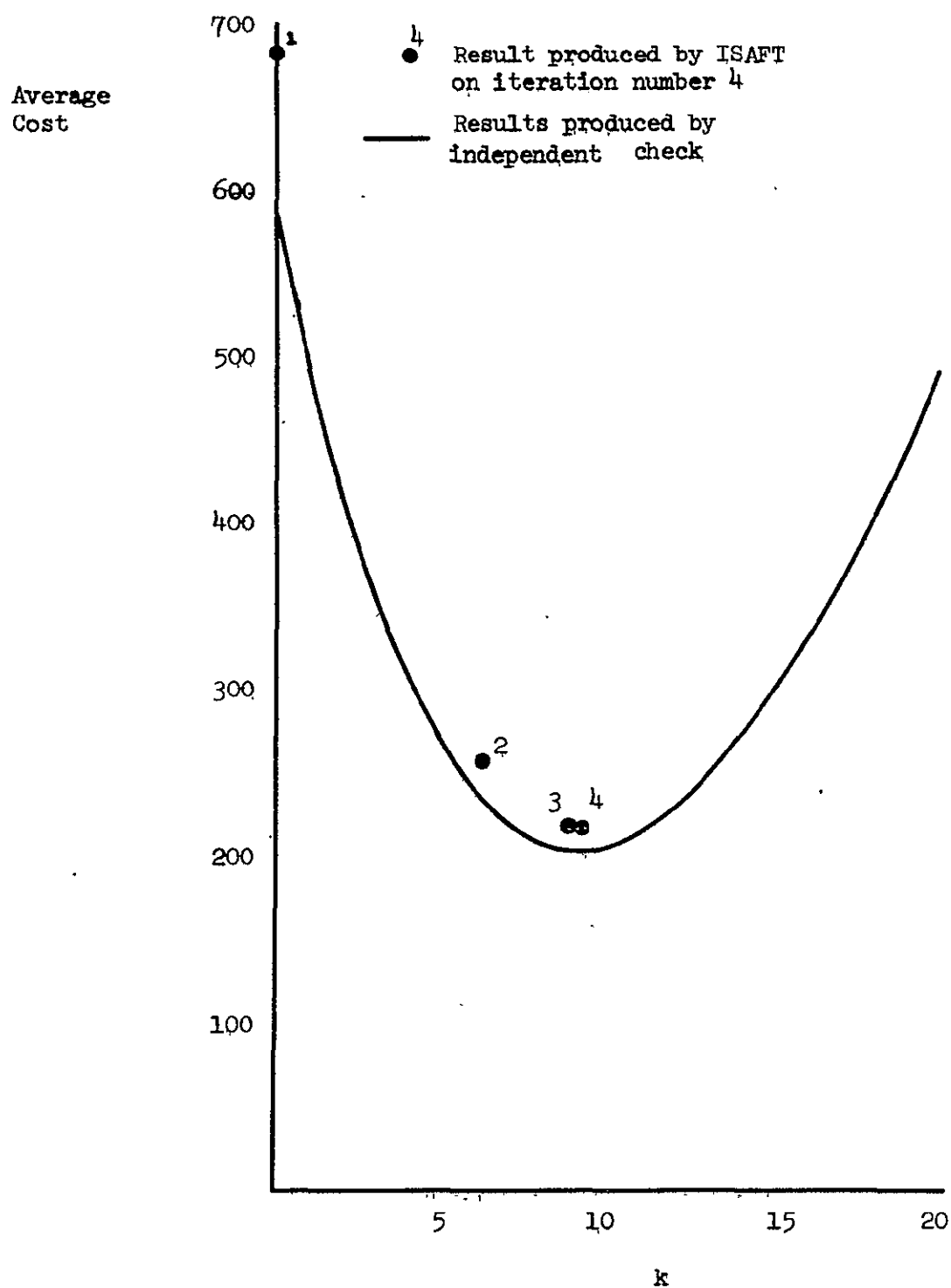


FIGURE III.5-7 Verification of the ISAFI results
for Example 3

satisfying certain necessary conditions. The necessary conditions are that the partial derivatives of the gain function with respect to each feedback gain should be zero. These necessary conditions may be satisfied by local maxima, minima or points of inflection. With a multidimensional problem points of inflection seldom occur in practice. The convergence factor μ built into the digital computer program ISAFIT prevents convergence to a local maximum. There still remains the problem of convergence to a local rather than a global minimum. No solution to this problem was found. It is felt, however, that it should be feasible to guarantee the existence of a single (global) minimum for suitable system and cost matrices, and useful work could be done in this area.

The digital computer program ISAFIT computes the various required functions of the state transition matrix (see (III.3-33)) through (III.3-38)) analytically. It was felt that this best illustrated the procedure. For high order systems however, it should be more economical to evaluate the state transition matrix at a number of discrete time instants and then evaluate the required functions by numerical integration.

CHAPTER IV

SUMMARY AND CONCLUSIONS

This report has considered the determination of optimal time-invariant output-feedback controllers for linear dynamic systems with quadratic cost functionals. The need for such controllers in practical engineering was illustrated, and the contributions of earlier researchers in the field were reviewed.

It was determined that deficiencies existed in two areas. For the case where optimization was to take place over the semi-infinite time interval Levine¹⁹ and Cassidy²⁰ had each derived a suitable computational algorithm. Both algorithms, however, required initialization by suitably stabilizing feedback gains and neither author gave a method for determination of such gains. For the case of optimization over a finite time interval, no satisfactory existing techniques were uncovered.

The problem of the determination of stabilizing feedback gains was approached via a gradient technique. The technique evolved from an eigenvalue sensitivity relationship given in Fadeev and Fadeeva²⁷. It was mechanized in the digital computer program GRADGN (Appendix I) and provides a practical method for determination of local stability maxima in feedback gain space. It is felt that this gradient technique should complement the existing optimization algorithms for the semi-infinite time interval problem.

A new technique was derived for the finite time interval problem. The technique is based on the Initial State Averaging concept, previously used for somewhat different problems by Levine¹⁹ and by Kleinman, Fortmann, and Athans²⁵. A computational algorithm was derived and is incorporated in the digital computer program ISAFIT (Appendix II). The algorithm satisfies a

set of necessary conditions by Newton-Raphson iteration, using their gradient with respect to the feedback gains. This is equivalent to minimization of the expected value of the cost functional by the method of second variations.

The contributions made by this report should aid in the search for practical optimal controllers.

Several outstanding problems remain. Foremost among these is the question of the sufficiency of the solutions obtained by the optimization technique described above. The technique was designed to determine local minima of the expected value of the cost functional. An examination of the convexity of this quantity in feedback gain space might uncover conditions ensuring a single (global) minimum. The gain initialization technique is similarly local and would also benefit by extension to a global technique.

The computational methods used in the digital computer program ISAFIT were designed to illustrate the theory. They are satisfactory for low order systems, but not for high order systems. It is felt that relatively simple modifications to ISAFIT should eliminate this deficiency.

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APPENDIX I

This Appendix comprises a listing of the gain initialization digital computer program GRADGN.

```

/JOB          4257          MCRKINN,LINES=55
1  DIMENSION W(3,4),XR(3),XI(3),VR(3),VI(3),IROW(3,2),
   1  GRAD(3,3),GRADI(3,3),GRAD(3,3),VRN(3),VIN(3),
   2  G(3,3)
2  COMMON AMAT(3,3),BMAT(3,3),AHAT(3,3),AAAA(90),ASUR(3,3),
   1  RR(3),RI(3),IANA(3)
   1  KPAL K(3,3)
   10  FORMAT (7I1J)
   20  FORMAT (4E13,/)
   25  FORMAT (20A4)
   30  FORMAT (7F10,4)
   35  FORMAT ('1',T4,'STATES',4X,'CONTROLS',4X,'FEEDBACKS',4X,'GAINS',
   1  4X,'IDELR',4X,'ISTOP')
   36  FORMAT (//T3,'STABILITY INCREASE INCREMENT =',F20.8)
1  37  FORMAT (//T3,'MINIMUM STABILITY REQUIRED FOR TERMINATION',3X,F20.8,
   1)
11  40  FORMAT (//T3,'SYSTEM MATRIX AMAT')
12  45  FORMAT (//T3,'CONTROL MATRIX BMAT')
13  50  FORMAT (//T2,'MAXIMUM REAL PART OF ROOTS')
14  60  FORMAT (//T3,'EIGVEC ERROR MESSAGES')
15  65  FORMAT (T3,'SOL= ',E10.4,10X,'ITER= ',I5,10X,'CIF= ',E10.4)
16  70  FORMAT (//T3,'EIGENVECTORS CORRES TO MRP EIGENVALUE')
17  75  FORMAT (T6,'ROW REAL PART',5X,'ROW IMAG PART',5X,'COL REAL PART',
   1  5X,'COL IMAG PART')
18  80  FORMAT (//T3,'GRADIENT MATRIX')
19  82  FORMAT ('1',T3,'PROGRAM TERMINATION DUE TO EXCESSIVE NUMBER OF ITE
   1RATIONS')
20  84  FORMAT ('1',T3,'SPECIFIED STABILITY HAS BEEN ATTAINED')
21  86  FORMAT ('1',T3,'GAIN SEARCH PROCEEDS ALONG A NEW GRADIENT. ITERATI
   1ON NUMBER',3X,I3)
22  87  FORMAT (//T3,'GAINS TOO FAR FROM STARTING POINT. COMPUTE NEW GRADI
   1ENT')
23  88  FORMAT (//T3,'STABILITY INCREASE STEP SIZE =',F20.8)
24  90  FORMAT (//T3,'STEP SIZE IS DOUBLED')
25  92  FORMAT (//T3,'CONVEX CURVE FIT TO STABILITY LOCUS')
26  93  FORMAT (//T3,'STEP SIZE TOO LARGE. RETURN TO PREVIOUS BEST GAINS')
27  94  FORMAT (//T3,'CURVE FIT NOT SUCCESSFUL. RETURN TO PREVIOUS BEST GA
   1INS')
28  95  FORMAT (//T3,'STEP SIZE TOO LARGE. COMPUTE NEW GRADIENT AT PREVIOUS
   1S BEST GAINS')
29  96  FORMAT ('1',T3,'STABILITY IMPROVEMENT RATE TOO SLOW. PROGRAM TERMI
   1NATED')
30  97  FORMAT (//T3,'LAST RESULTS SHOW MOST STABLE CONDITION FOUND')
31  98  FORMAT (//T3,'STEP SIZE TOO LARGE. PROGRAM CONTINUES WITH REDUCED
   1STEP SIZE')
C
C      NS = NUMBER OF SYSTEM STATES
C      NC = NUMBER OF CONTROLS
C      NF = NUMBER OF FEEDBACK STATES
C      IGAINS = 0 - INITIAL FEEDBACK GAINS ARE ZERO
C      = 1 - INITIAL FEEDBACK GAINS ARE READ IN
C      IDELR = 0 - STABILITY INCREMENT = 0.1
C      = 1 - STABILITY INCREMENT IS READ IN AS DELR
C      ISTOP = 0 - PROGRAM MAXIMIZES SYSTEM STABILITY
C      = 1 - PROGRAM TERMINATES AT STABILITY SPECIFIED BY STOPR

```

NOT REPRODUCIBLE

```

      C
32      READ (1,17) NS,NC,NF,IGAINS,IDE LR,ISTOP
33      WRITE (3,32)
34      WRITE (3,13) NS,NC,NF,IGAINS,IDE LR,ISTOP
      C
      C      DE LR CONTROLS STABILITY INCREASE INCREMENT
      C
35      IF (IDE LR) 120,120,110
36      110 READ (1,30) DE LR
37      GO TO 130
38      120 DE LR=0.1
39      130 DE LRI=DE LR
40      WRITE (3,36) DE LR
      C
      C      STOPR CONTROLS STABILITY REQUIRED FOR PROGRAM TERMINATION
      C
41      IF (ISTOP) 150,150,140
42      140 READ (1,30) STOPR
43      GO TO 160
44      150 STOPR=-1.0E6
45      160 WRITE (3,37) STOPR
46      DO 170 I=1,NS
47      DO 170 J=1,NC
48      170 K(I,J)=0.
      C
      C      SYSTEM DATA MATRICES ARE READ IN BY ROWS
      C      AMAT IS SYSTEM MATRIX
      C      BMAT IS CONTROL MATRIX
      C
49      READ (1,30) ((AMAT(I,J),J=1,NS),I=1,NS)
50      WRITE (3,40)
51      WRITE (3,20) ((AMAT(I,J),J=1,NS),I=1,NS)
52      READ (1,30) ((BMAT(I,J),J=1,NC),I=1,NS)
53      WRITE (3,45)
54      WRITE (3,20) ((BMAT(I,J),J=1,NC),I=1,NS)
      C
      C      INITIALIZATION OF GAIN MATRIX K
      C
55      IF (IGAINS) 190,190,180
56      180 READ (1,30) ((K(I,J),J=1,NC),I=1,NF)
57      190 KOUNT=1
58      ROOTT=0.
      C
      C      SUBROUTINE STAB COMPUTES THE SYSTEM STABILITY
      C
59      195 CALL STAB(NS,NC,NF,K)
60      198 IF(RR(1)-STOPR) 999,200,200
61      200 KOUNT=KOUNT+1
62      IF (KOUNT-1) 210,210,110
      C
      C      GAIN SEARCH PROCEEDS ALONG A NEW GRADIENT
      C
63      210 ROOTR=RR(1)
64      ROOTI=RI(1)
65      WRITE (3,53)

```

```

66      WRITE (3,2) ROOTR
67      RTEST=ROOTT
68      ROOTT=ROOTR
69      KI=KOUNT-1
70      WRITE (3,86) KI
      C
      C      COMPUTATION OF EIGENVECTORS
      C
71      CALL MSQ(NS)
72      CALL EIGVEC(3, AHAT, ASQR, W, IROW, XR, XI, VR, VI, ROOTR, ROOTI, NS, 3, 0,
      1SW1, ITER, DIF, 2)
      C
      C      SW1 = 1 FOR AN EXACT EIGENVALUE AND NO ROUND-OFF ERROR
      C      ITER = NUMBER OF ITERATIONS USED TO FIND EIGENVECTORS.
      C      IF TOLERANCE IS NOT ACHIEVED, PROGRAM ACCEPTS VALUES AT ITER = 15.
      C      DIF = LARGEST CHANGE IN ANY EIGENVECTOR COMPONENT AT FINAL ITER.
      C
73      WRITE (3,60)
74      WRITE (3,65) SW1, ITER, DIF
75      WRITE (3,7)
76      WRITE (3,75)
77      WRITE (3,20) (VR(I), VI(I), XR(I), XI(I), I=1, NS)
      C
      C      NORMALISE EIGENVECTORS INNER PRODUCT
      C
78      VECMGR=0.
79      VECMGI=0.
80      DO 240 I=1, NS
81          VECMGR=VECMGR+VR(I)*XR(I)-VI(I)*XI(I)
82      240  VECMGI=VECMGI+VR(I)*XI(I)+VI(I)*XR(I)
83          VECMGS=VECMGR*VECMGR+VECMGI*VECMGI
84      DO 250 I=1, NS
85          VRN(I)=(VR(I)*VECMGR+VI(I)*VECMGI)/VECMGS
86      250  VIN(I)=(VI(I)*VECMGR-VR(I)*VECMGI)/VECMGS
      C
      C      COMPUTE GRADIENT MATRIX
      C
87      DO 300 J=1, NF
88      DO 300 L=1, NC
89          GRADR(J,L)=0.
90          GRADI(J,L)=0.
91      DO 290 I=1, NS
92          GRADR(J,L)=GRADR(J,L)-VRN(I)*BMAT(I,L)
93      290  GRADI(J,L)=GRADI(J,L)+VIN(I)*BMAT(I,L)
94      300  GRAD(J,L)=GRADR(J,L)*XR(J)+GRADI(J,L)*XI(J)
95      WRITE (3,80)
96      WRITE (3,20) ((GRAD(I,J), J=1, NC), I=1, NF)
97      WRITE (3,88) DELR
98      GRDSQ=0.
99      DO 320 I=1, NF
100     DO 320 J=1, NC
101     320  GRDSQ=GRDSQ+GRAD(I,J)*GRAD(I,J)
102     DELK1=-DELR/GRDSQ
103     DO 340 I=1, NF
104     DO 340 J=1, NC

```

NOT REPRODUCIBLE


```

105      G(I,J)=K(I,J)
106      340  K(I,J)=G(I,J)+DELK1*GRAD(I,J)
107      ROOT1=ROOT3
108      CALL STAB(NS,NC,NF,K)
109      IF (RR(1)-STOPR) 999,390,390
110      390  RCOT2=RR(1)
111      KOUNT1=1
112      KOUNT2=1
113      IF (ROOT2-ROOT1) 400,470,620
114      400  IF((ROOT1-ROOT2)/DELK1-.95) 420,420,500
115      420  DELK2=.5*DELK1/(ROOT2-ROOT1+DELK1)
116      WRITE (3,92)
117      DO 440 I=1,NF
118      DO 440 J=1,NC
119      440  K(I,J)=G(I,J)+DELK2*GRAD(I,J)
120      CALL STAB(NS,NC,NF,K)
121      IF(RR(1)-STOPR) 999,999,460
122      460  ROOT3=RR(1)
123      IF(ROOT3-RCOT2) 470,470,480
124      470  IF(KOUNT-2) 475,475,472
125      472  IF(RTEST-RCOT3-DELK1) 1200,475,475
126      475  DELR=.5*(ROOT1-RCOT3)
127      GO TO 200
128      480  IF(KOUNT-2) 485,485,482
129      482  IF(RTEST-RCOT2-DELK1) 483,485,485
130      483  DO 484 I=1,NF
131      DO 484 J=1,NC
132      484  K(I,J)=G(I,J)+DELK1*GRAD(I,J)
133      WRITE (3,93)
134      CALL STAB(NS,NC,NF,K)
135      GO TO 1200
136      485  DELR=.5*(ROOT1-RCOT2)
137      WRITE (3,94)
138      DO 490 I=1,NF
139      DO 490 J=1,NC
140      490  K(I,J)=G(I,J)+DELK1*GRAD(I,J)
141      GO TO 195
142      500  DELK2=2.*DELK1
143      KOUNT1=KOUNT1+1
144      IF(KOUNT1-11) 505,505,502
145      502  DELR=.5*(ROOT1-RCOT2)
146      WRITE (3,87)
147      GO TO 200
148      505  WRITE (3,90)
149      DO 510 I=1,NF
150      DO 510 J=1,NC
151      510  K(I,J)=G(I,J)+DELK2*GRAD(I,J)
152      CALL STAB(NS,NC,NF,K)
153      IF(RR(1)-STOPR) 999,999,520
154      520  ROOT3=RR(1)
155      IF (ROOT3-ROOT2) 530,530,580
156      530  ROOT2=ROOT3
157      DELK1=DELK2
158      GO TO 400
159      580  IF(KOUNT-2) 585,585,582

```

```

160 582 IF(RTEST-ROOT2-DELR1) 583,585,585
161 583 DO 584 I=1,NF
162 DO 584 J=1,NC
163 584 K(I,J)=G(I,J)+DELK1*GRAD(I,J)
164 WRITE (3,93)
165 CALL STAB(NS,NC,NF,K)
166 GO TO 1200
167 585 DELR=.5*(ROOT1-RCOT2)
168 WRITE (3,95)
169 DO 590 I=1,NF
170 DO 590 J=1,NC
171 590 K(I,J)=G(I,J)+DELK1*GRAD(I,J)
172 GO TO 195
173 600 KOUNT2=KOUNT2+1
174 IF(KOUNT2-6) 605,605,1100
175 605 DELK2=.5*DELK1*(ROOT2-ROOT1+DELR)
176 WRITE (3,92)
177 DO 620 I=1,NF
178 DO 620 J=1,NC
179 620 K(I,J)=G(I,J)+DELK2*GRAD(I,J)
180 CALL STAB(NS,NC,NF,K)
181 IF(RR(1)-STOPA) 999,999,640
182 640 ROOT3=RR(1)
183 IF(ROOT3-ROOT1) 680,680,660
184 660 DELK1=DELK2
185 WRITE (3,98)
186 ROOT2=ROOT3
187 GO TO 600
188 680 IF(KOUNT-2) 685,685,682
189 682 IF(RTEST-ROOT3-DELR1) 1200,685,685
190 685 DELR=.5*(ROOT1-RCOT3)
191 GO TO 200
192 999 WRITE (3,84)
193 GO TO 2000
194 1100 WRITE (3,82)
195 GO TO 2000
196 1200 WRITE (3,96)
197 WRITE (3,97)
198 2000 CONTINUE
199 STOP
200 END

```

```

201      SUBROUTINE STAB(NS,NC,NF,K)
202      COMMON AMAT(30,30),EMAT(30,30),AHAT(30,30),AAAA(900),ASQR(30,30),
1      RR(30),RI(30),IANA(30)
203      REAL K(30,30)
204      10  FORMAT (//T3,'GAIN MATRIX K')
205      20  FORMAT (4E18.7)
206      30  FORMAT (//T2,'ROOTS',5X,'REAL PART',11X,'IMAG. PART')
207      40  FORMAT (2E2),0)
208      IA=NS
209      WRITE (3,10)
210      WRITE (3,20) ((K(I,J),J=1,NC),I=1,NF)
211      CALL AMHT(NS,NC,K)
212      CALL VECT(NS)
213      CALL HSRG(NS,AAAA,IA)
214      CALL ATEIG(NS,AAAA,RR,RI,IANA,IAL)
215      WRITE (3,30)
216      WRITE (3,40) (RR(I),RI(I),I=1,NS)
217      CALL MAXRT(NS)
218      RETURN
219      END

```

```

221      SUBROUTINE AMHT(NS,NC,K)
      C
      C      COMPUTES AHAT = AMAT - BMAT*K(TRANSPOSE)
      C
221      COMMON AMAT(30,30),BMAT(30,30),AHAT(30,30),AAAA(900),ASQR(30,30),
      1      RR(30),RI(30),IANA(30)
222      REAL K(30,30)
223      DO 100 I=1,VS
224      DO 100 J=1,NS
225      AHAT(I,J)=AMAT(I,J)
226      DO 100 L=1,NC
227      100 AHAT(I,J)=AHAT(I,J)-BMAT(I,L)*K(J,L)
228      RETURN
229      END

```

```

231      SUBROUTINE VECT(NS)
      C
      C   CONVERTS AHAT TO SINGLE SUBSCRIPT FORM AAAA.
      C
231      COMMON AMAT(30,30),BMAT(30,30),AHAT(30,30),AAAA(900),ASQR(30,30),
      1        BR(30),RI(30),IANA(30)
232      DO 100 J=1,NS
233      DO 100 I=1,NS
234      K=(J-1)*NS+I
235      100 AAAA(K)=AHAT(I,J)
236      RETURN
237      END

```

NOT REPRODUCIBLE

```

238      SUBROUTINE MSQ(NS)
      C
      C      COMPUTES ASQR = AHAT*AHAT
      C
239      COMMON AMAT(30,30),BMAT(30,30),AHAT(30,30),AAAA(900),ASQR(30,30),
      1      RR(30),RI(30),IANA(30)
240      DO 100 I=1,NS
241      DO 100 J=1,NS
242      ASQR(I,J)=0.
243      DO 100 K=1,NS
244      100 ASQR(I,J)=ASQR(I,J)+AHAT(I,K)*AHAT(K,J)
245      RETURN
246      END

```

```

249      SUBROUTINE HSDG(N,A,IA)
      C
      C  CONVERTS A TO UPPER HESSENBERG FORM
      C
250      DOUBLE PRECISION DABS,DFLOAT,DSIGN,DBLE,DEXP,DLOG,DLOG10,DATAN
      1,DSIN,DCOS,DSQRT,DTANH,DMOD,DMAX1,DMIN1
251      DOUBLE PRECISION S
252      DIMENSION A(9,J)
253      L=N
254      NIA=L*IA
255      LIA=NIA-IA
256      20 IF(L-3) 360,40,40
257      40 LIA=LIA-IA
258      L1=L-1
259      L2=L1-1
260      ISUB=LIA+L
261      IPIV=ISUB-IA
262      PIV=ABS(A(IPIV))
263      IF(L-3) 90,90,50
264      50 M=IPIV-IA
265      DO 80 I=L,M,IA
266      T=ABS(A(I))
267      IF(T-PIV) 80,80,60
268      60 IPIV=I
269      PIV=T
270      80 CONTINUE
271      90 IF(PIV) 100,320,100
272      100 IF(PIV-ABS(A(ISUB))) 180,180,120
273      120 M=IPIV-L
274      DO 140 I=1,L
275      J=M+I
276      T=A(J)
277      K=LIA+I
278      A(J)=A(K)
279      140 A(K)=T
280      M=L2-M/IA
281      DO 160 I=L1,NIA,IA
282      T=A(I)
283      J=I-M
284      A(I)=A(J)
285      160 A(J)=T
286      180 DO 200 I=L,LIA,IA
287      200 A(I)=A(I)/A(ISUB)
288      J=-IA
289      DO 240 I=1,L2
290      J=J+IA
291      LJ=L+J
292      DO 220 K=1,L1
293      KJ=K+J
294      KL=K+LIA
295      220 A(KJ)=A(KJ)-A(LJ)*A(KL)
296      240 CONTINUE
297      K=-IA
298      DO 300 I=1,N
299      K=K+IA
300

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NOT REPRODUCIBLE

```
300      LK=K+L1
301      S=A(LK)
302      LJ=L-IA
303      DO 280 J=1,L2
304      JK=K+J
305      LJ=LJ+IA
306      280 S=S+A(LJ)*A(JK)*1.0D5
307      300 A(LK)=S
308      DO 310 I=L,LIA,IA
309      310 A(I)=0.0
310      320 L=L1
311      GO TO 20
312      360 RETURN
313      END
```



```

314 SUBROUTINE ATEIG(M,A,RR,RI,IANA,IA)
      C
      C - COMPUTES ROOTS OF UPPER HESSENBERG MATRIX A
      C
315 DOUBLE PRECISION DABS,DLOAT,DSIGN,DBLE,DEXP,DLOG,DLOG10,DATAN
      P,DSIN,DCOS,DSQRT,DTANH,DMOD,DMAX1,DMIN1
316 DIMENSION A(900),RR(30),RI(30),PRR(2),PRI(2),IANA(30)
317 INTEGER P,P1,Q
318 B7=1.0E-8
319 E6=1.0E-6
320 E10=1.0E-10
321 DELTA=0.5
322 MAXIT=30
323 N=M
324 20 N1=N-1
325 IN=N1+IA
326 NN=IN+N
327 IF(N1) 30,1300,30
328 30 NP=N+1
329 IT=0
330 DO 40 I=1,2
331 PRR(I)=0.0
332 40 PRI(I)=0.0
333 PAN=0.0
334 PAN1=0.0
335 R=0.0
336 S=0.0
337 N2=N1-1
338 IN1=IN-IA
339 NN1=NN1+N
340 N1N=IN+N1
341 N1N1=IN1+N1
342 60 T=A(N1N1)-A(NN)
343 U=T*T
344 V=4.0*A(N1N)*A(NN1)
345 IF(ABS(V)-U*E7) 100,100,65
346 65 T=U+V
347 IF(ABS(T)-AMAX1(U,ABS(V))*E6) 67,67,68
348 67 T=0.0
349 68 U=(A(N1N1)+A(NN))/2.0
350 V=SQRT(ABS(T)/2.0)
351 IF(T) 70,70,70
352 70 IF(U) 80,75,75
353 75 RR(N1)=U+V
354 RR(N)=U-V
355 GO TO 130
356 80 RR(N1)=U-V
357 RR(N)=U+V
358 GO TO 130
359 100 IF(T) 120,110,110
360 110 RR(N1)=A(N1N1)
361 RR(N)=A(NN)
362 GO TO 130
363 120 RR(N1)=A(NN)
364 RR(N)=A(N1N1)

```

```

365      130 RI(N)=1.0
366      RI(N1)=1.0
367      GO TO 160
368      140 RR(N1)=L
369      RR(N)=U
370      RI(N1)=V
371      RI(N)=-V
372      160 IF(N2)1280,1230,180
373      180 N1N2=N1N1-IA
374      RMOD=RR(N1)*RR(N1)+RI(N1)*RI(N1)
375      EPS=E10*SQRT(RMOD)
376      IF(ABS(A(N1N2))-EPS)1280,1280,240
377      240 IF(ABS(A(N1N1))-E10*ABS(A(NN1)))1300,1300,250
378      250 IF(ABS(PAN1-A(N1N2))-ABS(A(N1N2))*E6)1240,1240,260
379      260 IF(ABS(PAN-A(NN1))-ABS(A(NN1))*E6)1240,1240,300
380      300 IF(IT-MAXIT)320,1240,1240
381      320 J=1
382      DO 360 J=1,2
383      K=NP-I
384      IF(ABS(RR(K)-PRR(I))+ABS(RI(K)-PRI(I))-DELTA*(ABS(RR(K))
385      +ABS(RI(K))))340,360,360
386      340 J=J+1
387      360 CONTINUE
388      GO TO (440,460,460,480),J
389      440 R=0.0
390      S=0.0
391      GO TO 500
392      460 J=N+2-J
393      R=RR(J)+RR(J)
394      S=RR(J)+RR(J)
395      GO TO 500
396      480 R=RR(N)*RR(N1)-RI(N)*RI(N1)
397      S=RR(N1)+RR(N1)
398      500 PAN=A(NN1)
399      PAN1=A(N1N2)
400      DO 520 J=1,2
401      K=NP-I
402      PRR(I)=RR(K)
403      PRI(I)=RI(K)
404      P=N2
405      IF(N-3)600,600,525
406      525 IPI=N1N2
407      DO 580 J=2,N2
408      IPI=IPI-IA-1
409      IF(ABS(A(IPI))-EPS)620,600,530
410      530 IPIP=IPI+IA
411      IPIP2=IPIP+IA
412      D=A(IPIP)*A(IPIP)-S1+A(IPIP2)*A(IPIP+1)+R
413      IF(D)540,560,540
414      540 IF(ABS(A(IPI)*A(IPIP+1))*(ABS(A(IPIP)+A(IPIP2+1)-S)+ABS(A(IPIP2+2)
415      +ABS(D)*EPS)620,620,560
416      560 P=N1-J
417      580 CONTINUE
418      600 Q=P
419      GO TO 680

```

```

418      620 P1=P-1
419      Q=P1
420      IF(P1-1)680,630,650
421      650 DO 660 I=2,P1
422      IPI=IPI-IA-1
423      IF(ABS(A(IPI))-EPS)680,680,660
424      660 Q=Q-1
425      680 II=(P-1)*IA+P
426      DO 1220 I=P,N1
427      III=II-IA
428      IIP=II+IA
429      IF(I-P)720,720,720
430      700 IPI=II+1
431      IPIP=IIP+1
432      G1=A(III)*(A(III)-S)+A(IIP)*A(IPI)+R
433      G2=A(IPI)*(A(IPIP)+A(III)-S)
434      G3=A(IPI)*A(IPIP+1)
435      A(IPI+1)=0.0
436      GO TO 780
437      720 G1=A(III)
438      G2=A(III+1)
439      IF(I-N2)740,740,760
440      740 G3=A(III+2)
441      GO TO 780
442      760 G3=0.0
443      780 CAP=SQRT(G1*G1+G2*G2+G3*G3)
444      IF(CAP)800,850,800
445      800 IF(G1)820,840,840
446      820 CAP=-CAP
447      840 T=G1+CAP
448      PSI1=G2/T
449      PSI2=G3/T
450      ALPHA=2.0/(1.0+PSI1*PSI1+PSI2*PSI2)
451      GO TO 880
452      860 ALPHA=2.0
453      PSI1=0.0
454      PSI2=0.0
455      880 IF(I-Q)900,960,900
456      900 IF(I-P)920,940,920
457      920 A(III)=-CAP
458      GO TO 960
459      940 A(III)=-A(III)
460      960 IJ=II
461      DO 1040 J=I,N
462      T=PSI1*A(IJ+1)
463      IF(I-N1)980,1000,1000
464      980 IP2J=IJ+2
465      T=T+PSI2*A(IP2J)
466      1000 ETA=ALPHA*(T+A(IJ))
467      A(IJ)=A(IJ)-ETA
468      A(IJ+1)=A(IJ+1)-PSI1*ETA
469      IF(I-N1)1020,1040,1040
470      1020 A(IP2J)=A(IP2J)-PSI2*ETA
471      1040 IJ=IJ+IA
472      IF(I-N1)1080,1060,1060

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NOT REPRODUCIBLE

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473      1060 K=N
474      GO TO 1100
475      1080 K=I+2
476      1100 IP=IIP-I
477      DO 1180 J=Q,K
478      JIP=IP+J
479      JI=JIP-IA
480      T=PSI1*A(JIP)
481      IF(I-N1)1120,1140,1140
482      1120 JIP2=JIP+IA
483      T=T+PSI2*A(JIP2)
484      1140 ETA=ALPHA*(T+A(JI))
485      A(JI)=A(JI)-ETA
486      A(JIP)=A(JIP)-ETA*PSI1
487      IF(I-N1)1160,1180,1180
488      1160 A(JIP2)=A(JIP2)-ETA*PSI2
489      1180 CONTINUE
490      IF(I-N2)1200,1220,1220
491      1200 JI=JI+3
492      JIP=JI+IA
493      JIP2=JIP+IA
494      ETA=ALPHA*PSI2*A(JIP2)
495      A(JI)=-ETA
496      A(JIP)=-ETA*PSI1
497      A(JIP2)=A(JIP2)-ETA*PSI2
498      1220 II=IIP+1
499      IT=IT+1
500      GO TO 60
501      1240 IF(ABS(A(NN1))-ABS(A(N1N2))) 1300,1280,1280
502      1280 IANA(N)=0
503      IANA(N1)=2
504      N=N2
505      IF(N2)1400,1400,20
506      1300 RR(N)=A(NN)
507      RI(N)=0.0
508      IANA(N)=1
509      IF(N1)1400,1400,1320
510      1320 N=N1
511      GO TO 20
512      1400 RETURN
513      END

```

```

514      SUBROUTINE MAXRT(NS)
      C
      C      COMPUTES ROOT HAVING MAXIMUM REAL PART.
      C
515      COMMON AMAT(30,30),BMAT(30,30),AHAT(30,30),AAAA(900),ASQR(30,30),
      1      RR(30),RI(30),IANA(30)
516      DO 100 I=2,NS
517      IF (RR(1)-RR(I))50,100,100
518      50  RI(I)=RI(I)
519      RR(I)=RR(I)
520      100 CONTINUE
521      RETURN
522      END

```

```

523      SUBROUTINE EIGVEC(IVC, A, B, N, IRDN, XR, XI, VR, VI, ROOTRE, ESX1
1      ROOTIE, NE, NMAX, T2, SW1, COUNTS, ERR, MMM) ESX1
C      SUBROUTINE TO FIND THE EIGENVECTORS OF A NON-SYMMETRIC MATRIX ESX1
C      BY A MODIFIED WILKINSON'S INVERSE ITERATION METHOD. ESX1
C      CONTROL IVC CODE IS ESX1
C      1 FIND ONLY THE REGULAR EIGENVECTORS (A X = LAMBDA X) ESX1
C      2 FIND ONLY THE TRANSPOSED EIGENVECTORS (AT V = LAMBDA V) ESX1
C      3 FIND BOTH TYPES OF EIGENVECTORS. ESX1
524      DIMENSION A(30,30), B(30,30), W(30,4), XR(30), XI(30), VR(30), VI(30),
1 IROW(30,2),
525      INTEGER COUNT, COUNTS, T2 ESX1
526      IQ1=1 ESX1
527      IQ3=3 ESX1
528      ROOTR = ROOTRE ESX1
529      ROOTI = ROOTIE ESX1
530      N = NE ESX1
531      MM = MMM - 1 ESX1
532      N1 = N - 1 ESX1
533      NP1 = N + 1 ESX1
534      IVC1 = IVC - 1 ESX1
535      IVC2 = IVC1 - 1 ESX1
536      COUNT = 1 ESX1
537      DO 400 I=1,N ESX1
538      W(I,1)=0.0E0 ESX1
539      XR(I)=0.0E0 ESX1
540      400 CONTINUE ESX1
541      CLIM = 1.0E-4 ESX1
542      IF(ROOTI) 1, 60, 1 ESX1
C      COMPLEX EIGENVALUE. ESX1
C      1 TEMP = - ROOTR - ROOTR ESX1
544      ISW = 2 ESX1
545      TEMP2=ROOTR*ROOTR+ROOTI*ROOTI ESX1
546      JJ = 300 ESX1
547      DO 606 I = 1, N ESX1
548      IF(T2) 600, 603, 600 ESX1
549      600 DO 602 J = 1, N ESX1
550      JJ = JJ + 1 ESX1
551      IF(JJ) 251, 602, 601, 601 ESX1
552      601 JJ = 1 ESX1
553      READ (T2) (W(I,1), W(I,2), W(I,3), W(I,4)) ESX1
554      602 B(I,J) = A(I,J)*TEMP + W(JJ,1) ESX1
555      GO TO 605 ESX1
556      603 DO 604 J = 1, N ESX1
557      604 B(I,J) = A(I,J)*TEMP + B(I,J) ESX1
558      605 B(I,J) = B(I,J) + TEMP2 ESX1
559      606 A(I,J) = A(I,J) - ROOTR ESX1
560      IF(T2 .NE. 0) REWIND T2 ESX1
561      GO TO 700 ESX1
562      607 IF(ICC) 622, 608, 622 ESX1
C      MATRIX SINGULAR. ESX1
C      622 IF(IVC2) 623, 625, 623 ESX1
563      622 IF(IVC2) 623, 625, 623 ESX1

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```

564      623 DO 624 LL = 1, N
565          W(LL,2)=0.0
566      624 XI(LL)=0.0
567          IF (VC1) 625, 514, 625
568      625 DO 626 LL = 1, N
569          W(LL,4)=0.0
570      626 VI(LL)=0.0
571          GO TO 511
C
C      MATRIX NOT SINGULAR.
C
572      608 DO 609 LL = 1, N
573          W(LL,1)=1.0
574          W(LL,2)=1.0
575          W(LL,3)=1.0
576      609 W(LL,4)=1.0
577      699 IF (VC2) 610, 612, 610
578      610 DO 611 I = 1, N
579          I2 = IROW(I,2)
580          XI(I2) = W(I,1)*ROOTI
581          DO 611 J = 1, N
582      611 XI(I2) = XI(I2) + A(I,J)*W(J,2)
583          IF (VC1) 612, 500, 612
584      612 DO 613 I = 1, N
585          VI(I) = W(I,3)*ROOTI
586          DO 613 J = 1, N
587      613 VI(I) = VI(I) + A(J,I)*W(J,4)
588          GO TO 499
589      615 CERR = 0.0
590          IF (VC2) 616, 619, 616
591      616 DO 618 I = 1, N
592          XR(I) = -W(I,2)
593          DO 617 J = 1, N
594      617 XR(I) = XR(I) + A(I,J)*XI(J)
595      618 XR(I) = XR(I)/ROOTI
596          IF (VC1) 619, 633, 619
597      619 DO 621 I = 1, N
598          VR(I) = -W(I,4)
599          DO 620 J = 1, N
600      620 VR(I) = VR(I) + A(J,I)*VI(J)
601      621 VR(I) = VR(I)/ROOTI
C
C      SEARCH VECTORS FOR LARGEST ELEMENT AND NORMALIZE.
C
602      627 AMAX = 0.0
603          DO 629 L = 1, N
604      629 TEMP = VR(L)**2 + VI(L)**2
605          IF (TEMP - AMAX) 629, 629, 628
606      628 AMAX = TEMP
607          I2 = L
608      629 CONTINUE
609      C1 = VR(I2)/AMAX
610      C2 = -VI(I2)/AMAX
611      DO 630 L = 1, N
612          TEMP = VI(L)

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613      VI(L) = VR(L)*C2 + TEMP*C1      ESY
614      630 VR(L) = VR(L)*C1 - TEMP*C2      ESY
615      IF(COUNT.EQ.1) GO TO 632      ESY
616      DO 631 LL = 1, N      ESY
617      631 CERR = AMAX1(CERR, ABS(VR(LL) - W(LL,3)), ABS(VI(LL) - W(LL,4)))      ESY
618      632 IF(IVC2) 633, 638, 633      ESY
619      633 AMAX = 0.0      ESY
620      DO 635 L = 1, N      ESY
621      TEMP = XR(L)**2 + XI(L)**2      ESY
622      IF(TEMP.EQ.0) 635, 635, 634      ESY
623      634 AMAX = TEMP      ESY
624      I2 = 1      ESY
625      635 CONTINUE      ESY
626      C1 = XR(I2)/AMAX      ESY
627      C2 = -XI(I2)/AMAX      ESY
628      DO 636 L = 1, N      ESY
629      TEMP = XI(L)      ESY
630      XI(L) = XR(L)*C2 + TEMP*C1      ESY
631      636 XR(L) = XR(L)*C1 - TEMP*C2      ESY
632      IF(COUNT.EQ.1) GO TO 646      ESY
633      DO 637 LL = 1, N      ESY
634      637 CERR = AMAX1(CERR, ABS(XR(LL) - W(LL,1)), ABS(XI(LL) - W(LL,2)))      ESY
635      C      ESY
636      C      TEST FOR CONVERGENCE.      ESY
637      C      ESY
638      638 IF(COUNT.EQ.1) GO TO 646      ESY
639      IF(CERR.GE.1.0E-4) GO TO 639      ESY
640      IF(CERR.GE.1.0E-8) GO TO 648      ESY
641      CLIM = CERR      ESY
642      IF(CLIM.GE.1.0E-8) GO TO 648      ESY
643      639 IF(COUNT.GE.15) GO TO 48      ESY
644      647 COUNT = COUNT + 1      ESY
645      IF(ROOT1) 642, 673, 642      ESY
646      642 IF(IVC2) 640, 644, 640      ESY
647      640 DO 641 LL = 1, N      ESY
648      W(LL,1) = XR(LL)      ESY
649      W(LL,2) = XI(LL)      ESY
650      IF(IVC1) 644, 610, 644      ESY
651      644 DO 645 LL = 1, N      ESY
652      W(LL,3) = VR(LL)      ESY
653      W(LL,4) = VI(LL)      ESY
654      GO TO 699      ESY
655      646 CERR = 0.0      ESY
656      IF(ICC) 648, 647, 648      ESY
657      648 ERR = CERR      ESY
658      COUNT = COUNT      ESY
659      IF(ROOT1) 667, 668, 667      ESY
660      667 DO 649 I = 1, N      ESY
661      649 A(I,1) = A(I,1) + ROOTR      ESY
662      RETURN      ESY
663      68 PRINT 101, ROOTR, ROOTI, CERR      ESY
664      GO TO 648      ESY
665      C      REAL EIGENVECTORS.      ESY
666      C      ESY

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662      60 ISW = 1
663      DO 651 I = 1, N
664      DO 650 J = 1, N
665      650 B(I,J) = A(I,J)
666      651 B(I,I) = B(I,I) - ROOTR
667      GO TO 700
668      652 IF (ICC) 680, 685, 680
669      C
670      C SINGULAR MATRIX.
671      680 IF (IVC2) 681, 683, 681
672      681 DO 682 L = 1, N
673      682 XI(L) = 0.0
674      IF (IVC1) 683, 514, 683
675      683 DO 684 L = 1, N
676      684 VI(L) = 0.0
677      GO TO 511
678      C
679      C MATRIX NOT SINGULAR.
680      C
681      685 IF (IVC2) 653, 656, 653
682      653 DO 654 L = 1, N
683      654 XI(L) = 1.0
684      IF (IVC1) 656, 500, 656
685      656 DO 657 L = 1, N
686      657 VI(L) = 1.0
687      GO TO 499
688      C
689      C NORMALIZE REAL VECTORS.
690      C
691      655 CERR = 0.0
692      IF (IVC2) 658, 662, 658
693      658 C1 = 0.0
694      C2 = 0.0
695      DO 660 L = 1, N
696      660 TEMP = ABS(XI(L))
697      IF (TEMP - C1) 660, 660, 659
698      659 C1 = TEMP
699      C2 = XI(L)
700      660 CONTINUE
701      DO 661 L = 1, N
702      661 XI(L) = XI(L)/C2
703      CERR = AMAX1(CERR, ABS(XI(L) - XR(L)))
704      661 XR(L) = XI(L)
705      IF (IVC1) 662, 638, 662
706      662 C2 = 0.0
707      C1 = 0.0
708      DO 664 L = 1, N
709      664 TEMP = ABS(VI(L))
710      IF (TEMP - C1) 664, 664, 663
711      663 C1 = TEMP
712      C2 = VI(L)
713      664 CONTINUE
714      DO 665 LL = 1, N
715      665 VI(LL) = VI(LL)/C2

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```

708      CERR = AMAX1(CERR, ABS(VI(LL) - W(LL,1)))      ESY
709      W(LL,1)=VI(LL)
710      665 VR(LL)=W(LL,1)
711      GO TO 638      ESY
712      668 IF(IVC2) 669, 671, 669      ESY
713      669 DO 670 I = 1, N      ESY
714      670 XI(I) = 0.0      ESY
715      IF(IVC1) 671, 70, 671      ESY
716      671 DO 672 I = 1, N      ESY
717      672 VI(I) = 0.0      ESY
718      70 RETURN      ESY
719      673 IF(IVC2) 674, 502, 674      ESY
720      674 DO 675 I = 1, N      ESY
721      I2 = IROW(I,2)      ESY
722      675 XI(I2) = XR(I)      ESY
723      GO TO 500      ESY
724      C      BACK SUBSTITUTION SECTION.      ESY
725      C      ESY
726      499 IF(IVC2) 500, 502, 500      ESY
727      500 DO 501 I = 2, N      ESY
728      I1 = I - 1      ESY
729      DO 501 J = 1, I1      ESY
730      501 XI(I) = XI(I) - B(I,J)*XI(J)      ESY
731      511 IF(IVC1) 502, 514, 502      ESY
732      502 DO 510 I = 1, N      ESY
733      I1 = I - 1      ESY
734      IF(I1) 503, 505, 503      ESY
735      503 DO 504 J = 1, I1      ESY
736      504 VI(I) = VI(I) - B(I,J)*VI(J)      ESY
737      IF(ICC) 505, 506, 505      ESY
738      505 IF(B(I,I)) 506, 507, 506      ESY
739      506 VI(I) = VI(I)/B(I,I)      ESY
740      GO TO 510      ESY
741      507 IF(VI(I)) 508, 509, 508      ESY
742      508 VI(I) = VI(I)*1.0E+15      ESY
743      GO TO 510      ESY
744      509 VI(I) = 1.0      ESY
745      510 CONTINUE      ESY
746      IF(IVC2) 514, 525, 514      ESY
747      514 DO 522 I = 1, N      ESY
748      IR = NP1 - I      ESY
749      IF(I - 1) 515, 517, 515      ESY
750      515 I2 = IR + 1      ESY
751      DO 516 J = I2, N      ESY
752      516 XI(IR) = XI(IR) - B(IR,J)*XI(J)      ESY
753      IF(ICC) 517, 518, 517      ESY
754      517 IF(B(IR,IR)) 518, 519, 518      ESY
755      518 XI(IR) = XI(IR)/B(IR,IR)      ESY
756      GO TO 522      ESY
757      519 IF(XI(IR)) 520, 521, 520      ESY
758      520 XI(IR) = XI(IR)*1.0E+15      ESY
759      GO TO 522      ESY
760      521 XI(IR) = 1.0      ESY
761      522 CONTINUE      ESY

```

```

760 IF(IVC1) 525, 529, 525 ESY
761 DO 526 I = 2, N ESY
762 IR = NP1 - 1 ESY
763 I2 = IR + 1 ESY
764 DO 526 J = I2, N ESY
765 526 VI(IR) = VI(IR) - B(J,IR)*VI(J) ESY
766 DO 527 L = 1, N ESY
767 I2 = IROW(L,1) ESY
768 527 VR(I2) = VI(L) ESY
769 DO 528 L = 1, N ESY
770 528 VI(L) = VR(L) ESY
771 529 IF(ROOT1) 615, 655, 615 ESY
C ESY
C FACTOR MATRIX. ESY
C ESY
772 700 ICC = 0 ESY
773 SW1 = 1.0E72 ESY
774 DO 701 LL = 1, N ESY
775 701 IROW(LL,1) = LL ESY
776 DO 708 K = 1, N1 ESY
777 AMAX = ABS(B(K,K)) ESY
778 IMAX = K ESY
779 K1 = K + 1 ESY
780 DO 702 I = K1, N ESY
781 IF(AMAX .GT. ABS(B(I,K))) GO TO 702 ESY
782 AMAX = ABS(B(I,K)) ESY
783 IMAX = I ESY
784 702 CONTINUE ESY
785 IF(AMAX .LT. SW1) SW1 = AMAX ESY
786 IF(AMAX .GE. 1.0E-25) GO TO 723 ESY
787 B(K,K) = 0.0 ESY
788 ICC = ICC + 1 ESY
789 GO TO 708 ESY
790 723 IF(IMAX .EQ. K) GO TO 704 ESY
791 DO 703 J = 1, N ESY
792 AMAX = B(K,J) ESY
793 B(K,J) = B(IMAX,J) ESY
794 703 B(IMAX,J) = AMAX ESY
795 I2 = IROW(K,1) ESY
796 IROW(K,1) = IROW(IMAX,1) ESY
797 IROW(IMAX,1) = I2 ESY
798 704 DO 707 I = K1, N ESY
799 IF(B(I,K)) 705, 707, 705 ESY
800 705 B(I,K) = B(I,K)/B(K,K) ESY
801 DO 706 J = K1, N ESY
802 706 B(I,J) = B(I,J) - B(K,J)*B(I,K) ESY
803 707 CONTINUE ESY
804 708 CONTINUE ESY
805 AMAX = ABS(B(N,N)) ESY
806 IF(AMAX - 1.0E-25) 712, 712, 713 ESY
807 712 B(N,N) = 0.0 ESY
808 SW1 = 0.0 ESY
809 ICC = ICC + 1 ESY
810 GO TO 709 ESY
811 713 IF(AMAX .LT. SW1) SW1 = AMAX ESY

```

```

812      759 IF(ICC .LE. 150) GO TO 710
813      IF(MM) 1051,1050,1051
814      1051 WRITE(103,102) ICC
815      COUNT = 0
816      RETURN
817      1050 WRITE(103,1052) ICC
818      710 DO 711 LL = 1, N
819      I2 = IROW(LL,1)
820      711 IROW(I2,2) = LL
821      IF(ROOT1) 607, 652, 607
822      1052 FORMAT(///23H ***** WARNING ***** , ' SUBROUTINE EIGVEC HAS
      1 FOUND AN EIGENVALUE OF APPARENT MULTIPLICITY',
      1 14, /23X, ' COMPUTATION OF EIGENVECTORS WILL CONTINUE AT USER'S OPTION'//)
823      101 FORMAT(38H) MORE THAN 15 LOOPS FOR EIGENVECTOR OF, 2E12.4,
      2 14H DIFFERENCE OF, E12.4)
824      102 FORMAT(16H ***** WARNING ***** , 14, 71H ZEROS ON DIAGONAL OF FACTORED
      1 MATRIX. CHECK FOR MULTIPLE EIGENVALUES, /20X,
      2 ' SUBROUTINE EIGVEC WILL NOT PERFORM COMPUTATION FOR THIS EIGENVALUE
      3 TOR '///)
825      END

```

/DATA

APPENDIX II

This Appendix comprises a listing of the I.S.A. optimization digital computer program ISAFI.

```

/JOB          4257      MCPRIAN,LINES=55
C
C      PROGRAM ISAPT
C      DETERMINES OPTIMAL CONSTANT GAIN OUTPUT FEEDBACK CONTROLLERS
C      FOR FINITE TIME STATE REGULATOR PROBLEMS
C      OUTPUTS ARE ASSUMED TO BE FIRST NF STATES
C
1      COMMON M(3,3),MI(3,3),RC(3),AHAT(3,3),T,K(3,3),G(3,3),GHAT(3,3),
1      FT(3,3),FFT1(3,3),DFT(3,3,3,3),FCFT1(3,3,3,3),
2      D2FT(3,3,3,3,3,3),DFDFT1(3,3,3,3,2,3),FC2FT1(3,3,3,3,3,3),
3      VW(9),VGRAC(9,9),GRADI(9,9),B(3,3),F(3,3),R(3,3),NS,NC,NF
2      DIMENSION A(3,3),C(3,3)
3      COMPLEX*16 M,MI,RC
4      DOUBLE PRECISION AHAT,T,K,G,GHAT,FT,FFT1,DFT,FCFT1,DFDFT1,
1      FC2FT1,VW,VGRAC,GRADI
5      DOUBLE PRECISION GF,GF2,TEST,TEST1,GASTCP
6      DOUBLE PRECISION CMAX1,DABS
7      10  FORMAT (4F14.7)
8      11  FORMAT (4D17.7)
9      15  FORMAT (7I11)
10     20  FORMAT (14I5)
11     25  FORMAT ('1',T4,'STATES',4X,'CONTROLS',4X,'FEEDBACKS',4X,'GAINS')
12     30  FORMAT (//T3,'INVERSE GRADIENT MATRIX')
13     35  FORMAT (//T3,'SYSTEM MATRIX A')
14     40  FORMAT (//T3,'NEW GAINS')
15     45  FORMAT (//T3,'CONTROL MATRIX B')
16     50  FORMAT (//T3,'GAIN MATRIX')
17     55  FORMAT (//T3,'TERMINAL TIME T =',F18.7)
18     60  FORMAT (//T7,'GAIN TOLERANCE ACHIEVED =',F13.7)
19     62  FORMAT (//T3,'AVERAGE COST =',D2.7)
20     65  FORMAT (//T3,'TERMINAL COST MATRIX F')
21     70  FORMAT (//T3,'REQUIRED STOPPING TOLERANCE =',F13.7)
22     75  FORMAT (//T3,'STATE WEIGHTING MATRIX Q')
23     80  FORMAT (//T3,'CONTROL WEIGHTING MATRIX R')
24     85  FORMAT ('1','ITERATION NUMBER',I11)
25     90  FORMAT (//T3,'SOLUTION IS COMPLETE. ABOVE GAINS ARE OPTIMAL')
26     92  FORMAT (//T3,'STEP SIZE TOO LARGE. NEW AVERAGE COST WOULD HAVE BEEN
1N',D16.7)
27     93  FORMAT (T3,'GAIN ADJUSTMENT IS HALVED')
28     95  FORMAT (//T3,'CONVERGENCE TOO SLOW. PROGRAM TERMINATED')
29     100  READ (1,20) NS,NC,NF,IGAINS
30     READ (1,10) ((A(I,J),J=1,NS),I=1,NS)
31     READ (1,10) ((B(I,J),J=1,NC),I=1,NS)
32     READ (1,11) T
33     READ (1,10) ((F(I,J),J=1,NS),I=1,NS)
34     READ (1,10) ((G(I,J),J=1,NS),I=1,NS)
35     READ (1,10) ((R(I,J),J=1,NC),I=1,NC)
36     READ (1,11) GASTCP
37     WRITE (3,25)
38     WRITE (3,15) NS,NC,NF,IGAINS
39     WRITE (3,35)
40     WRITE (3,10) ((A(I,J),J=1,NS),I=1,NS)
41     WRITE (3,45)
42     WRITE (3,10) ((B(I,J),J=1,NC),I=1,NS)
43     WRITE (3,55) T

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```

44      WRITE (3,65)
45      WRITE (3,1) ((F(I,J),J=1,NS),I=1,NS)
46      WRITE (3,75)
47      WRITE (3,1) ((Q(I,J),J=1,NS),I=1,NS)
48      WRITE (3,80)
49      WRITE (3,1) ((R(I,J),J=1,NC),I=1,NC)
50      NFP1=NF+1
51      NFC=NF*NC
52      DO 1000 I=1,NS
53      DO 1000 J=1,NC
54      G(I,J)=C.000
55      1000 K(I,J)=C.000
56      IF(IGAIN) 1200,1250,1100
57      1100 READ (1,11) ((K(I,J),J=1,NC),I=1,NF)
58      1200 CONTINUE
59      DO 1220 I=1,NS
60      DO 1220 J=NF,NS
61      QHAT(I,J)=C(I,J)
62      1220 AHAT(I,J)=A(I,J)
63      IT=1
64      WRITE (3,85) IT
65      WRITE (3,50)
66      WRITE (3,1) ((K(I,J),J=1,NC),I=1,NF)
67      1230 CONTINUE
68      DO 1250 I=1,NS
69      DO 1250 J=1,NF
70      AHAT(I,J)=A(I,J)
71      QHAT(I,J)=C(I,J)
72      DO 1250 N1=1,NC
73      AHAT(I,J)=AHAT(I,J)-P(I,N1)*K(J,N1)
74      DO 1250 N2=1,NC
75      1250 QHAT(I,J)=QHAT(I,J)+K(I,N1)*R(N1,N2)*K(J,N2)
76      CALL STRAN
77      1260 CALL FEFA
78      IT=IT+1
79      CALL GAINFN(GF)
80      WRITE (3,62) GF
81      CALL NESCON
82      CALL GRADNT
83      CALL INVERT(VGRAD,GRADI,NFC)
84      WRITE (3,30)
85      WRITE (3,1) ((GRADI(I,J),J=1,NFC),I=1,NFC)
86      CALL NEWKIT
87      WRITE (3,85) IT
88      WRITE (3,40)
89      WRITE (3,1) ((G(I,J),J=1,NC),I=1,NF)
90      IT1=C
91      1280 CONTINUE
92      DO 1300 I=1,NS
93      DO 1300 J=1,NF
94      AHAT(I,J)=A(I,J)
95      QHAT(I,J)=C(I,J)
96      DO 1300 N1=1,NC
97      AHAT(I,J)=AHAT(I,J)-B(I,N1)*G(J,N1)
98      DO 1300 N2=1,NC

```

```

99      1300 CHAT(I,J)=GFAT(I,J)+G(I,N1)*F(N1,N2)*G(J,N2)
100      CALL STRAP
101      CALL GAIN2(GF2)
102      IF(GF-GF2.GT..000000) GC TC 1380
103      WRITE (3,92) GF2
104      WRITE (3,93)
105      DO 1350 I=1,NF
106      DO 1350 J=1,NC
107      1350 G(I,J)=G(I,J)/2.+K(I,J)/2.
108      WRITE (3,40)
109      WRITE (3,10) ((G(I,J),J=1,NC),I=1,NF)
110      IT1=IT1+1
111      IF(IT1.LE.5) GC TC 1280
112      GO TC 3000
113      1380 TEST1=.000000
114      DO 2000 I=1,NF
115      DO 2000 J=1,NC
116      IF(G(I,J).EQ..000000) GC TC 1400
117      TEST=DABS((G(I,J)-K(I,J))/G(I,J))
118      TEST1=DMAX1(TEST,TEST1)
119      GC TC 2000
120      1400 IF(G(I,J)-K(I,J).EQ..000000) GC TC 2000
121      TEST1=10000.*GNSTCP
122      2000 K(I,J)=G(I,J)
123      WRITE (3,60) TEST1
124      WRITE (3,70) GNSTCP
125      IF(TEST1-GNSTCP.GT..000000) GC TC 1260
126      WRITE (3,90)
127      WRITE (3,62) GF2
128      GO TC 8000
129      3000 WRITE (3,95)
130      8000 CONTINUE
131      9000 CONTINUE
132      STOP
133      END

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NOT REPRODUCIBLE


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134      SUBROUTINE STRAM
      C
      C      COMPLETES THE STATE TRANSITION MATRIX
      C
135      COMMON M(3,3),MI(3,3),RC(3),AHAT(3,3),T,K(3,3),G(3,3),QHAT(3,3),
      1      FT(3,3),FFTI(3,3),DFT(3,3,3,3),FDFTI(3,3,3,3),
      2      D2FT(3,3,3,3,3,3),DFDFTI(3,3,3,3,3,3),FD2FTI(3,3,3,3,3,3),
      3      VW(9),VGRAD(9,9),GRADI(9,9),B(3,3),F(3,3),R(3,3),NS,NC,NF
136      DIMENSION AAAA(9),RR(3),RI(3),ASQR(3,3),ASQ2(3,3),XR(3),XI(3),VR(3
      1      ),VI(3),IANA(3),IRCW(3,2),VRN(3),VIN(3),W(3,4)
137      COMPLEX*16 M,MI,RC
138      COMPLEX*16 DCMPLEX,DCCNJG
139      DOUBLE PRECISION AHAT,T,K,G,CHAT,FT,FFTI,CFT,FDFTI,D2FT,DFDFTI,
      1      FD2FTI,VW,VGRAD,GRADI
140      DOUBLE PRECISION AAAA,RR,RI,ASQR,ASQ2,XR,XI,VR,VI,VRN,VIN,W,VECMGR
      1      ,VECMGI,VECMGS,SW1
141      10  FORMAT (2D18.7)
142      30  FORMAT (//T3,'SYSTEM EIGENVALUES')
143      CALL VECT(AHAT,AAAA,NS)
144      CALL HSBG(NS,AAAA,NS)
145      CALL ATEIG(NS,AAAA,RR,RI,IANA,NS)
146      WRITE (3,30)
147      WRITE (3,10) (RR(I),RI(I),I=1,NS)
148      CALL MSC(AAAA,NS,ASQR)
149      DO 100 I=1,NS
150      RC(I)=DCMPLEX(RR(I),RI(I))
151      DO 100 J=1,NS
152      ASQ2(I,J)=ASQR(I,J)
153      CALL EIGVEC(3,AHAT,ASQR,W,IRCW,XR,XI,VR,VI,RR(1),RI(1),NS,NS,0,
      1      SW1,ITER,DIF,2)
154      VECMGR=C.OOO
155      VECMGI=C.OOO
156      DO 110 I=1,NS
157      VECMGR=VECMGR+VR(I)*XR(I)-VI(I)*XI(I)
158      110  VECMGI=VECMGI+VR(I)*XI(I)+VI(I)*XR(I)
159      VECMGS=VECMGR*VECMGR+VECMGI*VECMGI
160      DO 120 I=1,NS
161      VRN(I)=(VR(I)*VECMGR+VI(I)*VECMGI)/VECMGS
162      VIN(I)=(VI(I)*VECMGR-VR(I)*VECMGI)/VECMGS
163      M(I,I)=DCMPLEX(XR(I),XI(I))
164      120  MI(I,I)=DCMPLEX(VRN(I),VIN(I))
165      DO 100 KOUNT=2,NS
166      KOUNT1=KOUNT-1
167      IF(RR(KOUNT)-RR(KOUNT1)) 200,140,200
168      140  CONTINUE
169      DO 150 I=1,NS
170      M(I,KOUNT)=CCCNJG(M(I,KOUNT1))
171      150  MI(KOUNT,I)=DCCNJG(MI(KOUNT1,I))
172      GO TO 100
173      200  DO 210 I=1,NS
174      210  DO 210 J=1,NS
175      210  ASQR(I,J)=ASQ2(I,J)
176      CALL EIGVEC(3,AHAT,ASQR,W,IRCW,XR,XI,VR,VI,RR(KOUNT),RI(KOUNT),NS,
      1      NS,0,SW1,ITER,DIF,2)
177      VECMGR=C.OOO

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```

178      VECMGI=C.CC.
179      DO 220 I=1,NS
180      VECMGR=VECMGR+VR(I)*XR(I)-VI(I)*XI(I)
181      220 VECMGI=VECMGI+VR(I)*XI(I)+VI(I)*XR(I)
182      VECMGS=VECMGR*VECMGR+VECMGI*VECMGI
183      DO 230 I=1,NS
184      VRN(I)=(VR(I)*VECMGR+VI(I)*VECMGI)/VECMGS
185      VIN(I)=(VI(I)*VECMGR-VR(I)*VECMGI)/VECMGS
186      M(I,KOUNT)=CCPLX(XR(I),XI(I))
187      230 MI(KOUNT,I)=CCPLX(VRN(I),VIN(I))
188      1000 CONTINUE
189      RETURN
190      END

```

```

191      SUBROUTINE FEFA
C
C      COMPUTES REQUIRED FUNCTIONS OF THE STATE TRANSITION MATRIX
C
192      COMMON M(3,3),MI(3,3),RC(3),AHAT(3,3),T,K(3,3),G(3,3),UHAT(3,3),
1          FT(3,3),FFTI(3,3),CFT(3,3,3,3),FCFTI(3,3,3,3),
2          D2FT(3,3,3,3,3,3),CDFDFTI(3,3,3,3,3,3),FC2FTI(3,3,3,3,3,3),
3          VW(9),VGRAD(9,9),GRADI(9,9),B(3,3),F(3,3),R(3,3),NS,NC,RF
193      DIMENSION D1(3,3,3,3,3,3),D2(3,3,3,3,3,3)
194      COMPLEX*16 EX1(3),EX2(3,3),R1(3,3),R2(3,3),DUM1,DUM2,DUM3,DUM4,
1          DUM5,DUM6,DUM7
195      COMPLEX*16 M,MI,RC
196      COMPLEX*16 CDEXP
197      DOUBLE PRECISION AHAT,T,K,G,CHAT,FT,FFTI,CFT,FCFTI,D2FT,CDFDFTI,
1          FC2FTI,VW,VGRAD,GRADI
198      DOUBLE PRECISION D1,D2
199      NS1=NS-1
200      NF1=NF-1
201      NC1=NC-1
202      DO 100 I=1,NS
203      EX1(I)=CDEXP(RC(I)*T)
204      DO 100 J=1,NS
205      EX2(I,J)=CDEXP((RC(I)+RC(J))*T)
206      R1(I,J)=RC(I)+RC(J)
207      R2(I,J)=RC(I)-RC(J)
208      DO 1000 I=1,NS
209      DO 1000 J=1,NS
210      FT(I,J)=C.DC0
211      FFTI(I,J)=C.DC0
212      DO 300 K1=1,NF
213      DO 300 K2=1,NC
214      CFT(I,J,K1,K2)=C.DC0
215      FDFDFTI(I,J,K1,K2)=C.DC0
216      DO 300 K3=1,NF
217      DO 300 K4=1,NC
218      D2FT(I,J,K1,K2,K3,K4)=C.DC0
219      CDFDFTI(I,J,K1,K2,K3,K4)=C.DC0
220      FD2FTI(I,J,K1,K2,K3,K4)=C.DC0
221      DO 1000 N1=1,NS
222      FT(I,J)=FT(I,J)+M(I,N1)*MI(N1,J)*EX1(N1)
223      DO 1000 N2=1,NS
224      DO 1000 N3=1,NS
225      DUM1=M(I,N1)*MI(N1,N2)*M(J,N3)
226      FFTI(I,J)=FFTI(I,J)+DUM1*MI(N3,N2)*(EX2(N1,N3)-1.)/R1(I,J)
227      DO 600 K1=1,NF
228      DO 600 K2=1,NC
229      DUM2=M(I,N1)*MI(N1,N2)*B(N2,K2)*M(K1,N3)
230      IF(N1-N3) 320,310,320
231      310 CFT(I,J,K1,K2)=CFT(I,J,K1,K2)-DUM2*MI(N3,J)*T*EX1(N1)
232      GO TO 330
233      320 DFT(I,J,K1,K2)=DFT(I,J,K1,K2)-DUM2*MI(N3,J)*(EX1(N3)-EX1(N1))/R2(N1,N3)
234      330 DO 600 N4=1,NS
235      DO 600 N5=1,NS
236      DUM3=DUM1*MI(N3,N4)*B(N4,K2)*M(K1,N5)*MI(N5,N2)

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NOT REPRODUCIBLE

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237      IF(N3-N5) 350,340,350
238 340 FDFTI(I,J,K1,K2)=FCFTI(I,J,K1,K2)-DUM3*(EX2(N1,N3)*(R1(N1,N3)*T-1.
      1)+1.)/(R1(N1,N3)*R1(N1,N3))
239      GC TC 360
240 350 FDFTI(I,J,K1,K2)=FCFTI(I,J,K1,K2)-DUM3*((EX2(N1,N5)-1.)/R1(N1,N5)-
      1*(EX2(N1,N3)-1.)/R1(N1,N3))/R2(N5,N3)
241 360 DO 600 K3=1,N4
242      DO 600 K4=1,N4
243          DUM4=2.*DUM2*M(I(N3,N4)*B(N4,K4)*M(K3,N5)*M(I(N5,J)
244          IF(N3-N5) 400,370,400
245 370 IF(N1-N3) 390,380,390
246 380 C2FT(I,J,K1,K2,K3,K4)=C2FT(I,J,K1,K2,K3,K4)+DUM4*T*T*EX1(N1)/2.
247      GC TC 455
248 390 C2FT(I,J,K1,K2,K3,K4)=C2FT(I,J,K1,K2,K3,K4)+DUM4*((R2(N3,N1)*T-1.)
      1*EX1(N3)+EX1(N1))/(R2(N3,N1)*R2(N3,N1))
249      GC TC 455
250 400 CCNTINUE
251      IF(N1-N3) 420,410,420
252 410 C2FT(I,J,K1,K2,K3,K4)=C2FT(I,J,K1,K2,K3,K4)-DUM4*T*EX1(N1)/R2(N5,N
      1)
253      CO TC 430
254 420 C2FT(I,J,K1,K2,K3,K4)=C2FT(I,J,K1,K2,K3,K4)-DUM4*(EX1(N3)-EX1(N1)
      1/(R2(N3,N1)*R2(N5,N3))
255 430 IF(N1-N5) 440,450,440
256 440 C2FT(I,J,K1,K2,K3,K4)=C2FT(I,J,K1,K2,K3,K4)+DUM4*(EX1(N5)-EX1(N1)
      1/(R2(N5,N1)*R2(N5,N3))
257      GC TC 455
258 450 C2FT(I,J,K1,K2,K3,K4)=C2FT(I,J,K1,K2,K3,K4)+DUM4*T*EX1(N5)/R2(N5,N
      1)
259 455 DO 600 N6=1,N5
260      DO 600 N7=1,N5
261          DUM5=M(I(N4,N5)*B(N5,K4)*M(K3,N6)*M(I(N6,N7)
262          DUM6=DUM2*M(I(N3,N7)*M(J,N4)*DUM5
263          DUM7=2.*M(I,N1)*M(I(N1,N7)*B(J,N2)*M(I(N2,N3)*B(N3,K2)*M(K1,N4)*DUM5
264          IF(N1-N3) 490,460,490
265 460 IF(N4-N6) 480,470,480
266 470 CFDFTI(I,J,K1,K2,K3,K4)=CFDFTI(I,J,K1,K2,K3,K4)+DUM6*((T*T*R1(N1,N
      14)*R1(N1,N4)-2.*T*R1(N1,N4)+2.)*EX2(N1,N4)-2.)/(R1(N1,N4)*R1(N1,N4
      2)*R1(N1,N4))
267      GC TC 520
268 480 CFDFTI(I,J,K1,K2,K3,K4)=CFDFTI(I,J,K1,K2,K3,K4)+DUM6*((R1(N1,N6)*
      1T-1.)*EX2(N1,N6)+1.)/(R1(N1,N6)*R1(N1,N6))-((R1(N1,N4)*T-1.)*EX2(N
      21,N4)+1.)/(R1(N1,N4)*R1(N1,N4))/R2(N6,N4)
269      GC TC 520
270 490 IF(N4-N6) 510,500,510
271 500 CFDFTI(I,J,K1,K2,K3,K4)=CFDFTI(I,J,K1,K2,K3,K4)+DUM6*((R1(N3,N4)*
      1T-1.)*EX2(N3,N4)+1.)/(R1(N3,N4)*R1(N3,N4))-((R1(N1,N4)*T-1.)*EX2(N
      21,N4)+1.)/(R1(N1,N4)*R1(N1,N4))/R2(N3,N1)
272      GC TC 520
273 510 CFDFTI(I,J,K1,K2,K3,K4)=CFDFTI(I,J,K1,K2,K3,K4)+DUM6*((EX2(N3,N6)-
      11.)/R1(N3,N6)-(EX2(N3,N4)-1.)/R1(N3,N4)-(EX2(N1,N6)-1.)/R1(N1,N6)+
      2*(EX2(N1,N4)-1.)/R1(N1,N4))/(R2(N3,N1)*R2(N6,N4))
274 520 IF(N4-N6) 560,530,560
275 530 IF(N2-N4) 550,540,550
276 540 FD2FTI(I,J,K1,K2,K3,K4)=FD2FTI(I,J,K1,K2,K3,K4)+.5*DUM7*(EX2(N1,N2

```

```

1)*(T*T*R1(N1,N2)*R1(N1,N2)-2.*T*R1(N1,N2)+2.)*-2.)/(R1(N1,N2)*R1(N1
2,N2)*R1(N1,N2))
277 GO TC 6C2
278 55G FD2FTI(I,J,K1,K2,K3,K4)=FD2FTI(I,J,K1,K2,K3,K4)+DUM7*((EX2(N1,N4)*
1(R1(N1,N4)*T-1.))+1.)/(R1(N1,N4)*R1(N1,N4))-EX2(N1,N4)-1.)/(R1(N1,
2N4)*R2(N4,N2))+EX2(N1,N2)-1.)/(R1(N1,N2)*R2(N4,N2))/R2(N4,N2)
279 GO TC 6C3
280 56C IF(N2-N4) 58C,57G,58C
281 57G FD2FTI(I,J,K1,K2,K3,K4)=FD2FTI(I,J,K1,K2,K3,K4)-DUM7*(EX2(N1,N4)*
1(R1(N1,N2)*T-1.))+1.)/(R1(N1,N2)*R1(N1,N2)*R2(N6,N2))
282 GO TC 59C
283 58G FD2FTI(I,J,K1,K2,K3,K4)=FD2FTI(I,J,K1,K2,K3,K4)-DUM7*((EX2(N1,N4)
1-1.)/R1(N1,N4)-(EX2(N1,N2)-1.)/R1(N1,N2))/(R2(N4,N2)*R2(N6,N4))
284 59C IF(N2-N6) 594,592,594
285 592 FD2FTI(I,J,K1,K2,K3,K4)=FD2FTI(I,J,K1,K2,K3,K4)+DUM7*(EX2(N1,N6)*
1(R1(N1,N6)*T-1.))+1.)/(R1(N1,N6)*R2(N6,N4)*R1(N1,N6))
286 GO TO 6C3
287 594 FD2FTI(I,J,K1,K2,K3,K4)=FD2FTI(I,J,K1,K2,K3,K4)+DUM7*((EX2(N1,N6)-
1.)/R1(N1,N6)-(EX2(N1,N2)-1.)/R1(N1,N2))/(R2(N6,N2)*R2(N6,N4))
288 6CG CONTINUE
289 DO 7C3 K1=1,NF
290 DO 7C3 K2=1,NC
291 DO 7C3 K3=1,NF
292 DO 7C3 K4=1,NC
293 D1(I,J,K1,K2,K3,K4)=D2FT(I,J,K1,K2,K3,K4)
294 73D D2(I,J,K1,K2,K3,K4)=FD2FTI(I,J,K1,K2,K3,K4)
295 DO 8C3 K1=1,NF
296 DO 8C3 K2=1,NC
297 DO 8C3 K3=1,NF
298 DO 8C3 K4=1,NC
299 D2FT(I,J,K1,K2,K3,K4)=(D1(I,J,K1,K2,K3,K4)+C1(I,J,K3,K4,K1,K2))/2.
300 80D FD2FTI(I,J,K1,K2,K3,K4)=(D2(I,J,K1,K2,K3,K4)+C2(I,J,K3,K4,K1,K2))/
12.
301 1CC CONTINUE
302 RETURN
303 END

```

```

304      SUBROUTINE SAINFA(CF)
      C
      C      COMPLETE AVERAGE COST
      C
305      COMMON N(3,3),MI(3,3),RC(3),AHAT(3,3),T,K(3,3),G(3,3),GHAT(3,3),
1          FT(3,3),FFTI(3,3),DFT(3,3,3,3),FCFTI(3,3,3,3),
2          D2FT(3,3,3,3,3,3),DCDFTI(3,3,3,3,2,3),FC2FTI(3,3,3,3,2,3),
3          VH(9),VGRAD(9,9),GRADI(9,9),P(3,3),F(3,3),K(3,3),NS,NC,NF
306      COMPLEX*16 N,MI,RC
307      DOUBLE PRECISION AHAT,T,K,G,CHAT,FI,FFTI,DFT,FLFTI,D2FT,LCDFTI,
1          FC2FTI,VH,VGRAD,GRADI
308      DOUBLE PRECISION GF
309      GF=C.OCC
310      DO 1000 N1=1,NS
311      DO 1000 N2=1,NS
312      GF=GF+GHAT(N1,N2)*FFTI(N2,N1)
313      DO 1000 N3=1,NS
314      1000 GF=GF+F(N1,N2)*FT(N2,N3)*FT(N1,N3)
315      GF=GF/NS
316      RETURN
317      END

```

NOT REPRODUCIBLE

```

318      SUBROUTINE NESCON
      C
      C      COMPUTES NECESSARY CONDITIONS
      C
319      COMMON P(3,3),MI(3,3),RC(3),AHAT(3,3),T,K(3,3),G(3,3),QHAT(3,3),
1         FT(3,3),FFTI(3,3),DFT(3,3,3,3),FCFTI(3,3,3,3),
2         D2FT(3,3,3,3,3,3),DFDFTI(3,3,3,3,3,3),FC2FTI(3,3,3,3,3,3),
3         VW(9),VGRAD(9,9),GRADI(9,9),B(3,3),F(3,3),R(3,3),NS,NC,NF
320      COMPLEX*16 P,MI,RC
321      DOUBLE PRECISION AHAT,T,K,G,CHAT,FT,FFTI,DFT,FCFTI,D2FT,DFDFTI,
1         FD2FTI,VW,VGRAD,GRADI
322      10  FORMAT (//T3,'NECESSARY CONDITIONS VECTOR')
323      20  FORMAT (4D18.7)
324      NFC=NF*NC
325      DO 1100 I=1,NF
326      DO 1100 J=1,NC
327      IN=NF*(J-1)+I
328      VW(IN)=0.000
329      DO 1100 N1=1,NS
330      DO 1000 N2=1,NS
331      VW(IN)=VW(IN)+QHAT(N1,N2)*FDFTI(N2,N1,I,J)
332      DO 1000 N3=1,NS
333      1000 VW(IN)=VW(IN)+F(N1,N2)*FT(N2,N3)*DFT(N1,N3,I,J)
334      DO 1100 N4=1,NC
335      1100 VW(IN)=VW(IN)+R(J,N4)*K(N1,N4)*FFTI(N1,I)
336      WRITE (3,10)
337      WRITE (3,20) (VW(I),I=1,NFC)
338      RETURN
339      END

```

```

340      SUBROUTINE GRADNT
      C
      C      COMPLETES GRADIENT MATRIX
      C
341      COMMON M(3,3),MI(3,3),RC(3),AHAT(3,3),T,K(3,3),G(3,3),CHAT(3,3),
1      FT(3,3),FFTI(3,3),DFT(3,3,3,3),FDFFTI(3,3,3,3),
2      D2FT(3,3,3,3,3,3),DFDFTI(3,3,3,3,3,3),FC2FTI(3,3,3,3,3,3),
3      VH(9),VGRAD(9,9),GRADI(9,9),R(3,3),F(3,3),R(3,3),NS,NC,NF
342      COMPLEX*16 M,MI,RC
343      DOUBLE PRECISION AHAT,T,K,G,CHAT,FT,FFTI,DFT,FDFFTI,D2FT,DFDFTI,
1      FC2FTI,VH,VGRAD,GRADI
344      10  FORMAT (//I3,'GRADIENT MATRIX')
345      20  FORMAT (4D18.7)
346      NFC=NF*NC
347      DO 1100 I=1,NF
348      DO 1100 J=1,NC
349      IA=NF*(J-1)+I
350      DO 1100 K1=1,NF
351      DO 1100 K2=1,NC
352      ID=NF*(K2-1)+K1
353      VGRAD(IA,IC)=R(J,K2)*FFTI(K1,I)
354      DO 1100 N1=1,NS
355      DO 1000 N2=1,NS
356      VGRAD(IN,IC)=VGRAD(IN,ID)+CHAT(N1,N2)*(DFDFTI(N2,N1,K1,K2,I,J)+FD2
1      IFTI(N2,N1,K1,K2,I,J))
357      DO 1000 N3=1,NS
358      1000 VGRAD(IN,ID)=VGRAD(IN,ID)+F(N1,N2)*(DFT(N2,N3,K1,K2)*CFT(N1,N3,I,J
2      )+FT(N2,N3)*D2FT(N1,N3,K1,K2,I,J))
359      DO 1100 N4=1,NC
360      1100 VGRAD(IN,IC)=VGRAD(IN,ID)+R(K2,N4)*K(N1,N4)*(FDFFTI(N1,K1,I,J)+
1      FDFFTI(K1,N1,I,J)) + R(J,N4)*K(N1,N4)*
2      (FDFFTI(N1,I,K1,K2)+FDFFTI(I,N1,K1,K2))
361      WRITE (3,10)
362      WRITE (3,20) ((VGRAD(I,J),J=1,NFC),I=1,NFC)
363      RETURN
364      END

```



```

365      SUBROUTINE INVERT(A,P,N)
      C
      C      INVERTS A TO GIVE B
      C
366      DIMENSION A(9,9),B(9,9)
367      DOUBLE PRECISION A,B,C,D,X
368      DOUBLE PRECISION DABS
369      IF(N-1) 103,100,101
370      100  B(1,1)=1.0D0/A(1,1)
371      RETURN
372      101  DO 102 I=1,N
373           DO 102 J=1,N
374           B(I,J)=0.000
375           DO 103 I=1,N
376           103  B(I,I)=1.0D0
      C      PICK UP PIVOT ELEMENT
377      DO 114 K=1,N
378      L=K
379      IF(N-K) 112,110,104
380      104  I=K+1
381      DO 106 JJ=I,N
382      IF(DABS(A(JJ,K))-DABS(A(L,K))) 106,106,105
383      105  L=JJ
384      106  CONTINUE
385      IF(L-K) 107,110,107
      C      PERFORM ROW INTERCHANGE
386      DO 108 J=K,N
387      C=A(K,J)
388      A(K,J)=A(L,J)
389      108  A(L,J)=C
390      DO 109 J=1,K
391      C=B(K,J)
392      B(K,J)=B(L,J)
393      109  B(L,J)=C
      C      COLUMN ELIMINATION
394      110  DO 114 I=1,N
395      IF(K-I) 111,114,111
396      111  D=A(I,K)/A(K,K)
397      DO 112 J=K,N
398      112  A(I,J)=A(I,J)-D*A(K,J)
399      A(I,K)=0.000
400      DO 113 J=1,N
401      113  B(I,J)=B(I,J)-D*B(K,J)
402      114  CONTINUE
      C      SOLVE FOR INVERSE
403      DO 115 J=1,N
404      DO 115 I=1,N
405      X=B(I,J)/A(I,I)
406      115  B(I,J)=X
407      RETURN
408      END

```

```

409      SUBROUTINE NEWKIT
      C
      C      PERFORMS NEWTON RAPHSON ITERATION
      C
410      COMMON M(3,3),MI(3,3),RC(3),AHAT(3,3),T,K(3,3),G(3,3),CHAT(3,3),
      1      FT(3,3),FFTI(3,3),DFT(3,3,3,3),FCFTI(3,3,3,3),
      2      C2FT(3,3,3,3,3,3),DFDFTI(3,3,3,3,3,3),FC2FTI(3,3,3,3,3,3),
      3      VW(9),VGRAD(9,9),GRADI(9,9),R(3,3),F(3,3),R(3,3),NS,NC,NF
411      COMPLEX*16 M,MI,RC
412      DOUBLE PRECISION AHAT,T,K,G,CHAT,FT,FFTI,CFT,FCFTI,C2FT,CFDFTI,
      1      FC2FTI,VW,VGRAD,GRADI
413      DO 1000 I=1,NF
414      DO 1000 J=1,NC
415      G(I,J)=K(I,J)
416      IN=NF*(J-1)+I
417      DO 1000 K1=1,NF
418      DO 1000 K2=1,NC
419      ID=NF*(K2-1)+K1
420      1000 G(I,J)=G(I,J)-GRADI(IN,ID)*VW(ID)
421      RETURN
422      END

```

```

423      SUBROUTINE VECT(AHAT,AAAA,NS)
      C
      C   CONVERTS AHAT TO SINGLE SUBSCRIPT FORM AAAA
      C
424      DIMENSION AHAT(3,3),AAAA(9)
425      DOUBLE PRECISION AHAT,AAAA
426      DO 100 J=1,NS
427      DO 100 I=1,NS
428      K=(J-1)*NS+I
429      100 AAAA(K)=AHAT(I,J)
430      RETURN
431      END

```

```

432      SUBROUTINE GAIN2(GF2)
      C
      C      COMPLETES AVERAGE COST FOR CONVERGENCE CHECK
      C
433      COMMON P(3,3),MI(3,3),RC(3),AHAT(3,3),T,K(3,3),G(3,3),QHAT(3,3),
1      FT(3,3),FFTI(3,3),DFT(3,3,3,3),FDFTI(3,3,3,3),
2      D2FT(3,3,3,3,3,3),DFDFTI(3,3,3,3,3,3),FC2FTI(3,3,3,3,3,3),
3      VW(9),VGRAD(9,9),GRADI(9,9),R(3,2),F(3,3),R(3,3),NS,NC,NF
434      COMPLEX*16 EX1(3),EX2(3,3),R1(3,3),DUM1
435      COMPLEX*16 M,MI,RC
436      COMPLEX*16 CDEXP
437      DOUBLE PRECISION AHAT,T,K,G,CHAT,FT,FFTI,DFT,FCFTI,D2FT,DFDFTI,
1      FC2FTI,VW,VGRAD,GRADI
438      DOUBLE PRECISION GF2
439      DO 500 I=1,NS
440      EX1(I)=CDEXP(RC(I)*T)
441      DO 500 J=1,NS
442      EX2(I,J)=CDEXP((RC(I)+RC(J))*T)
443      500 R1(I,J)=RC(I)+RC(J)
444      DO 1000 I=1,NS
445      DO 1000 J=1,NS
446      FT(I,J)=0.000
447      FFTI(I,J)=0.000
448      DO 1000 N1=1,NS
449      FT(I,J)=FT(I,J)+P(I,N1)*MI(N1,J)*EX1(N1)
450      DO 1000 N2=1,NS
451      DO 1000 N3=1,NS
452      DUM1=P(I,N1)*MI(N1,N2)*P(J,N3)
453      1000 FFTI(I,J)=FFTI(I,J)+DUM1*MI(N3,N2)*(EX2(N1,N3)-1.)/R1(N1,N3)
454      GF2=0.000
455      DO 2000 N1=1,NS
456      DO 2000 N2=1,NS
457      GF2=GF2+QHAT(N1,N2)*FFTI(N2,N1)
458      DO 2000 N3=1,NS
459      2000 GF2=GF2+F(N1,N2)*FT(N2,N3)*FT(N1,N3)
460      GF2=GF2/NS
461      RETURN
462      END

```

```

463      SUBROUTINE HSBG(N,A,IA)
      C
      C      CONVERTS A TO UPPER HESSENBERG FORM
      C
464      DIMENSION A(9)
465      DOUBLE PRECISION A,PIV,T,S
466      DOUBLE PRECISION DABS
467      L=N
468      NIA=L*IA
469      LIA=NIA-IA
470      20 IF(L-3) 360,40,40
471      40 LIA=LIA-IA
472      L1=L-1
473      L2=L1-1
474      ISUB=LIA+L
475      IPIV=ISUB-IA
476      PIV=DABS(A(IPIV))
477      IF(L-3) 90,90,50
478      50 M=IPIV-IA
479      DO 80 I=L,M,IA
480      T=DABS(A(I))
481      IF(T-PIV) 80,80,60
482      60 IPIV=I
483      PIV=T
484      80 CONTINUE
485      90 IF(PIV) 100,320,100
486      100 IF(PIV-DABS(A(ISUB))) 180,180,120
487      120 M=IPIV-L
488      DO 140 I=1,L
489      J=M+I
490      T=A(J)
491      K=LIA+I
492      A(J)=A(K)
493      140 A(K)=T
494      M=L2-M/IA
495      DO 160 I=L1,NIA,IA
496      T=A(I)
497      J=I-M
498      A(I)=A(J)
499      160 A(J)=T
500      180 DO 200 I=L,LIA,IA
501      200 A(I)=A(I)/A(ISUB)
502      J=-IA
503      DO 240 I=1,L2
504      J=J+IA
505      LJ=L+J
506      DO 220 K=1,L1
507      KJ=K+J
508      KL=K+LIA
509      220 A(KJ)=A(KJ)-A(LJ)*A(KL)
510      240 CCNTINUE
511      K=-IA
512      DO 300 I=1,N
513      K=K+IA
514      LK=K+L1

```

```
515      S=A(LK)
516      LJ=L-IA
517      DO 280 J=1,L2
518      JK=K+J
519      LJ=LJ+IA
520      S=S+A(LJ)*A(JK)*1.GDC
521      300 A(LK)=S
522      DO 310 I=L,LIA,IA
523      310 A(I)=0.GDC
524      320 L=L1
525      GO TC 20
526      360 RETURN
527      END
```

```

528      SUBROUTINE ATEIG(M,A,RR,RI,IANA,IA)
      C
      C      COMPUTES ROOTS OF UPPER HESSENBERG MATRIX A
      C
529      DIMENSION A(9),RR(3),RI(3),PRR(2),PRI(2),IANA(3)
530      DOUBLE PRECISION E7,E6,E10,DELTA,PRR,PRI,PAN,PANI,R,S,T,A,U,V,RR,
      RI,RMOD,EPS,D,G1,G2,G3,CAP,PSI1,PSI2,ALPHA,ETA
531      DOUBLE PRECISION DABS,DSQRT,DMAX1
532      INTEGER P,P1,Q
533      E7=1.00-8
534      E6=1.00-6
535      E10=1.00-10
536      DELTA=0.500
537      MAXIT=30
538      N=M
539      20 N1=N-1
540      IN=N1*IA
541      NN=IN*N
542      IF(N1) 30,1300,30
543      30 NP=N+1
544      IT=0
545      DO 40 I=1,2
546      PRR(I)=0.000
547      40 PRI(I)=0.000
548      PAN=0.000
549      PANI=0.000
550      R=0.000
551      S=0.000
552      N2=N1-1
553      IN1=IN-IA
554      NN1=IN1+N
555      N1N=IN+N1
556      N1N1=IN1+N1
557      60 T=A(N1N1)-A(NN)
558      U=T*T
559      V=4.000*A(N1N1)*A(NN)
560      IF(DABS(V)-U+E7) 100,100,65
561      65 T=U+V
562      IF(DABS(T)-DMAX1(U,DABS(V))*E6) 67,67,68
563      67 T=0.000
564      68 U=(A(N1N1)+A(NN))/2.000
565      V=DSQRT(DABS(T))/2.000
566      IF(T)140,70,70
567      70 IF(U) 80,75,75
568      75 RR(N1)=U+V
569      RR(N)=U-V
570      GO TO 130
571      80 RR(N1)=U-V
572      RR(N)=U+V
573      GO TO 130
574      100 IF(T)120,110,110
575      110 RR(N1)=A(N1N1)
576      RR(N)=A(NN)
577      GO TO 130
578      120 RR(N1)=A(NN)

```

```

579      RR(N)=A(N1N1)
580 130 RI(N)=3.CDC
581      RI(N1)=C.CCC
582      RI(N1)=C.CC
583      GO TC 160
584 140 RR(N1)=L
585      RR(N)=U
586      RI(N1)=V
587      RI(N)=~V
588 160 IF(N2)1280,1280,180
589 180 N1N2=N1N1-IA
590      RMOD=RR(N1)*RR(N1)+RI(N1)*RI(N1)
591      EPS=E1J*DSQRT(RMOD)
592      IF(DABS(A(N1N2))-EPS) 1280,1280,240
593 240 IF(DABS(A(N1N1))-E1C*DABS(A(N1N1))) 1300,1300,250
594 250 IF(DABS(PAN1-A(N1N2))-CABS(A(N1N2))*E6) 1240,1240,260
595 260 IF(DABS(PAN-A(N1N1))-CABS(A(N1N1))*E6) 1240,1240,300
596 300 IF(IT-MAXIT) 320,1240,1240
597 320 J=1
598      CC 360 I=1,2
599      K=NP-I
600      IF(DABS(RR(K)-PRR(I))+DABS(RI(K)-PRI(I))-DELTA*(DABS(RR(K))
1      +DABS(RI(K)))) 340,360,360
601 340 J=J+I
602 360 CCNTINUE
603      GC TC (440,460,460,480),J
604 440 R=3.CCC
605      S=0.CCC
606      GO TC 500
607 460 J=N+2-J
608      R=RR(J)*RR(J)
609      S=RR(J)+RR(J)
610      GO TC 500
611 480 R=RR(N)*RR(N1)-RI(N)*RI(N1)
612      S=RR(N)+RR(N1)
613 500 PAN=A(N1N1)
614      PAN1=A(N1N2)
615      CC 520 I=1,2
616      K=NP-I
617      PRR(I)=RR(K)
618 520 PRI(I)=RI(K)
619      P=N2
620      IF(N-3)600,600,525
621 525 IPI=N1N2
622      CC 560 J=2,N2
623      IPI=[IPI-IA-1
624      IF(DABS(A(IPI))-EPS) 600,600,530
625 530 IPIP=IPI+IA
626      IPIP2=IPIP+IA
627      C=A(IPIP)*(A(IPIP)-S)+A(IPIP2)*A(IPIP+1)+R
628      IF(D)540,560,540
629 540 IF(DABS(A(IPI)*A(IPIP+1))*(DABS(A(IPIP)+A(IPIP2+1)-S)+CABS(A(IPIP2
1      +2))) - CABS(C)*EPS) 620,620,560
630 560 P=N1-J
631 580 CCNTINUE

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NOT REPRODUCIBLE


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632      600 Q=P
633      GO TO 680
634      620 P1=P-1
635      Q=P1
636      IF(P1-1)680,680,650
637      650 DO 660 I=2,P1
638      IPI=IPI-IA-1
639      IF(DABS(A(IPI))-EPS) 680,680,660
640      660 Q=Q-1
641      680 II=(P-1)*IA+P
642      DO 1220 I=P,N1
643      I-I1=II-IA
644      IIP=II+IA
645      IF(I-P)720,700,720
646      700 IPI=II+1
647      IPIP=IIP+1
648      G1=A(II)*(A(II)-S)+A(IIP)*A(IPI)+R
649      G2=A(IPI)*(A(IPIP)+A(II)-S)
650      G3=A(IPI)*A(IPIP+1)
651      A(IPI+1)=C,G00
652      GO TO 780
653      720 G1=A(I11)
654      G2=A(I11+1)
655      IF(I-N2)740,740,760
656      740 G3=A(I11+2)
657      GO TO 780
658      760 G3=J,G00
659      CAP=CSQRT(G1*G1+G2*G2+G3*G3)
660      IF(CAP)800,860,800
661      800 IF(G1)820,840,840
662      820 CAP=-CAP
663      840 T=G1+CAP
664      PSI1=G2/T
665      PSI2=G3/T
666      ALPHA=2.000/(1.000+PSI1*PSI1+PSI2*PSI2)
667      GO TO 880
668      860 ALPHA=2.000
669      PSI1=0.000
670      PSI2=0.000
671      880 IF(I-Q)900,960,900
672      900 IF(I-P)920,940,920
673      920 A(I11)=-CAP
674      GO TO 960
675      940 A(I11)=-A(I11)
676      960 IJ=II
677      DO 1040 J=I,N1
678      T=PSI1*A(IJ+1)
679      IF(I-N1)980,1000,1000
680      980 IP2J=IJ+2
681      T=T+PSI2*A(IP2J)
682      1000 ETA=ALPHA*(T+A(IJ))
683      A(IJ)=A(IJ)-ETA
684      A(IJ+1)=A(IJ+1)-PSI1*ETA
685      IF(I-N1)1020,1040,1040
686      1020 A(IP2J)=A(IP2J)-PSI2*ETA

```

```

687      1040 IJ=IJ+IA
688          IF(I-N1)1080,1060,1060
689      1060 K=N
690          GO TO 1100
691      1080 K=I+2
692      1100 IP=IIP-I
693          DO 1180 J=Q,K
694              JIP=IP+J
695              JI=JIP-IA
696              I=PSI1*A(JIP)
697              IF(I-N1)1120,1140,1140
698      1120 JIP2=JIP+IA
699              I=I+PSI2*A(JIP2)
700      1140 ETA=ALPHA*(I+A(JI))
701              A(JI)=A(JI)-ETA
702              A(JIP)=A(JIP)-ETA*PSI1
703              IF(I-N1)1160,1180,1180
704      1160 A(JIP2)=A(JIP2)-ETA*PSI2
705      1180 CCNTINUE
706              IF(I-N2)1200,1220,1220
707      1200 JI=JI+3
708              JIP=JI+IA
709              JIP2=JIP+IA
710              ETA=ALPHA*PSI2*A(JIP2)
711              A(JI)=-ETA
712              A(JIP)=-ETA*PSI1
713              A(JIP2)=A(JIP2)-ETA*PSI2
714      1220 II=IIP+1
715              IT=IT+1
716              GO TO 60
717      1240 IF(DABS(A(NN1))-DABS(A(N1N2)))1300,1280,1280
718      1280 IANA(N)=0
719              IANA(N1)=2
720              N=N2
721              IF(N2)1400,1400,20
722      1300 RR(N)=A(NN)
723              RI(N)=0.000
724              IANA(N)=1
725              IF(N1)1400,1400,1320
726      1320 N=N1
727              GO TO 20
728      1400 RETURN
729          END

```

```

730      SUBROUTINE MSC(AAAA,NS,ASQR)
      C
      C      COMPLETES ASQR = AAAA*AAAA
      C
731      DIMENSION AAAA(9),ASQR(3,3)
732      DOUBLE PRECISION AAAA,ASQR
733      DO 100 I=1,NS
734      DO 100 J=1,NS
735      ASQR(I,J)=0.000
736      DO 100 K=1,NS
737      K1=NS*(J-1)+K
738      K2=NS*(K-1)+I
739      100 ASQR(I,J)=ASQR(I,J)+AAAA(K1)*AAAA(K2)
740      RETURN
741      END

```

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742      SUBROUTINE EIGVEC(IVC, A, B, W, IRCH, XR, XI, VR, VI, RCOTRE, CS
1      RCGTIE, NE, AMAX, T2, SW1, COUNT, ERR, MM) ES
C      SUBROUTINE TO FIND THE EIGENVECTORS OF A NON-SYMMETRIC MATRIX ES
C      BY A MODIFIED WILKINSON'S INVERSE ITERATION METHOD. ES
C      CONTROL IVC CODE IS ES
C      1 FIND ONLY THE REGULAR EIGENVECTORS (A X = LAMBDA X) ES
C      2 FIND ONLY THE TRANSPOSED EIGENVECTORS (AT V = LAMBDA V) ES
C      3 FIND BOTH TYPES OF EIGENVECTORS. ES
743      DIMENSION A(3,3),B(3,3),W(3,2),XR(3),XI(3),VR(3),VI(3),IRON(3,2)
744      DOUBLE PRECISION RCOTR,RCCTI,RCCTR,RCCTIE,TEMP,TEMP2,AMAX,C1,C2,
1      SW1,W,XR,XI,VR,VI,B,ZERR,ECERR,A
745      DOUBLE PRECISION DABS,CSIGN,CSQRT,DMAX1
746      INTEGER COUNT, COUNT, T2 ES
747      IO1=1
748      IO3=3
749      RCOTR = RCOTRE ES
750      RCOTI = RCCTIE ES
751      N = NE ES
752      MM = MM - 1 ES
753      N1 = N - 1 ES
754      NP1 = N + 1 ES
755      IVC1 = IVC - 1 ES
756      IVC2 = IVC1 - 1 ES
757      COUNT = 1 ES
758      DO 400 I=1,N
759      W(I,1)=G.CDC
760      XR(I)=0.CDC
761      400 CONTINUE
762      CLIM = 1.CE-4 ES
763      IF(RCOTI) 1, 63, 1 ES
C      COMPLEX EIGENVALUE. ES
C      1 TEMP = - RCOTR - RCCTR ES
764      ISW = 2 ES
765      TEMP2=RCOTR*RCCTR+RCCTI*RCCTI ES
766      JJ = 300 ES
767      DO 606 I = 1, N ES
768      IF(T2) 600, 603, 600 ES
769      600 DO 602 J = 1,N ES
770      JJ = JJ + 1 ES
771      IF(JJ - 251) 602, 601, 601 ES
772      601 JJ = 1 ES
773      READ (T2) (W(LL,1), LL = 1,250) ES
774      B(I,J) = A(I,J)*TEMP + W(JJ,1) ES
775      GO TO 605 ES
776      603 DO 604 J = 1, N ES
777      B(I,J) = A(I,J)*TEMP + B(I,J) ES
778      604 B(I,I) = B(I,I) + TEMP2 ES
779      605 A(I,I) = A(I,I) - RCCTR ES
780      IF(T2,NE,0) REWIND T2 ES
781      GC TC 700 ES
782      607 IF(ICC) 622, 608, 622 ES
783      C      MATRIX SINGULAR. ES
C

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      C
784      622 IF(IVC2) 623, 625, 623
785      623 DO 624 LL = 1, N
786          W(LL,2)=0.000
787      624 XI(LL)=0.000
788          IF(IVC1) 625, 514, 625
789      625 DO 626 LL = 1, N
790          W(LL,4)=0.000
791      626 VI(LL)=0.000
792          GO TO 511
      C
      C      MATRIX NOT SINGULAR.
      C
793      608 DO 609 LL = 1, N
794          W(LL,1)=1.000
795          W(LL,2)=1.000
796          W(LL,3)=1.000
797      609 W(LL,4)=1.000
798      699 IF(IVC2) 610, 612, 610
799      610 DO 611 I = 1, N
800          I2 = IRCW(I,2)
801          XI(I2) = W(I,1)*RCCTI
802          DO 611 J = 1, N
803      611 XI(I2) = XI(I2) + A(I,J)*W(J,2)
804          IF(IVC1) 612, 500, 612
805      612 DO 613 I = 1, N
806          VI(I) = W(I,3)*RCCTI
807          DO 613 J = 1, N
808      613 VI(I) = VI(I) + A(J,I)*W(J,4)
809          GO TO 499
810      615 CERR = 0.0
811          DCERR = J.GD3
812          IF(IVC2) 616, 619, 616
813      616 DO 618 I = 1, N
814          XR(I) = -W(I,2)
815          DO 617 J = 1, N
816      617 XR(I) = XR(I) + A(I,J)*XI(J)
817      618 XR(I) = XR(I)/RCCTI
818          IF(IVC1) 619, 633, 619
819      619 DO 621 I = 1, N
820          VR(I) = -W(I,4)
821          DO 620 J = 1, N
822      620 VR(I) = VR(I) + A(J,I)*VI(J)
823      621 VR(I) = VR(I)/RCCTI
      C
      C      SEARCH VECTORS FOR LARGEST ELEMENT AND NORMALIZE.
      C
824      627 AMAX=0.000
825      629 L = 1, N
826          TEMP = VR(L)**2 + VI(L)**2
827          IF(TEMP = AMAX) 629, 629, 628
828      628 AMAX = TEMP
829          I2 = L
830      629 CONTINUE
831          C1 = VR(I2)/AMAX

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832      C2 = -VI(I2)/AMAX
833      DO 630 L = 1, N
834      TEMP = VI(L)
835      VI(L) = VR(L)*C2 + TEMP*C1
836      630 VR(L) = VR(L)*C1 - TEMP*C2
837      IF(COUNT .EQ. 1) GO TO 632
838      DO 631 LL = 1, N
839      DCERR=DMAX1(DCERR,DABS(VR(LL)-W(LL,3)),DABS(VI(LL)-W(LL,4)))
840      632 IF(IVC2) 633, 636, 633
841      633 AMAX=J.DO
842      DO 635 L = 1, N
843      TEMP = XR(L)**2 + XI(L)**2
844      IF(TEMP - AMAX) 635, 635, 634
845      634 AMAX = TEMP
846      I2 = L
847      635 CONTINUE
848      C1 = XR(I2)/AMAX
849      C2 = -XI(I2)/AMAX
850      DO 636 L = 1, N
851      TEMP = XI(L)
852      XI(L) = XR(L)*C2 + TEMP*C1
853      636 XR(L) = XR(L)*C1 - TEMP*C2
854      IF(COUNT .EQ. 1) GO TO 646
855      DO 637 LL = 1, N
856      DCERR=DMAX1(DCERR,DABS(XR(LL)-W(LL,1)),DABS(XI(LL)-W(LL,2)))
C
C      TEST FOR CONVERGENCE.
C
857      638 IF(COUNT .EQ. 1) GO TO 646
858      CERR=DCERR
859      IF(CERR .GE. 1.0E-4) GO TO 639
860      IF(CERR .GE. CLIM) GO TO 648
861      CLIM = CERR
862      IF(CLIM .LE. 1.0E-8) GO TO 648
863      639 IF(COUNT .GE. 15) GO TO 68
864      647 COUNT = COUNT + 1
865      IF(RCOTI) 642, 673, 642
866      642 IF(IVC2) 642, 644, 642
867      645 DO 641 LL = 1, N
868      W(LL,1) = XR(LL)
869      641 W(LL,2) = XI(LL)
870      IF(IVC1) 644, 610, 644
871      644 DO 645 LL = 1, N
872      W(LL,3) = VR(LL)
873      645 W(LL,4) = VI(LL)
874      GO TO 659
875      646 CERR = J.J
876      DCERR=J.DO
877      IF(ICC) 648, 647, 648
878      648 ERR = CERR
879      COUNTC = COUNT
880      IF(RCOTI) 667, 668, 667
881      667 DO 649 I = 1, N
882      649 A(I,I) = A(I,I) + RCCTR
883      RETURN

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NOT REPRODUCIBLE


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927      IF(TEMP - C1) 664, 664, 663      ES\
928      663 C1 = TEMP                      ES\
929      C2 = VI(LL)                      ES\
930      664 CONTINUE                      ES\
931      DO 665 LL = 1, N                    ES\
932      VI(LL) = VI(LL)/C2                  ES\
933      DCERR=CMAX1(DCERR,CABS(VI(LL)-W(LL,1)))
934      W(LL,1)=VI(LL)
935      665 VR(LL)=W(LL,1)
936      GO TO 638                          ES\
937      668 IF(IVC2) 669, 671, 669          ES\
938      669 DO 670 L = 1, N                 ES\
939      670 XI(L)=0.000
940      IF(IVC1) 671, 70, 671              ES\
941      DO 672 L = 1, N                    ES\
942      672 VI(L)=0.000
943      70 RETURN                          ES\
944      673 IF(IVC2) 674, 502, 674          ES\
945      674 DO 675 I = 1, N                 ES\
946      I2 = IRCk(I,2)                     ES\
947      675 XI(I2) = XR(I)                 ES\
948      GO TO 500                          ES\
C
C      BACK SUBSTITUTION SECTION.
C
949      459 IF(IVC2) 500, 502, 500          ES\
950      DO 501 I = 2, N                    ES\
951      I1 = I - 1                          ES\
952      DO 501 J = 1, I1                   ES\
953      501 XI(I) = XI(I) - B(I,J)*XI(J)    ES\
954      511 IF(IVC1) 502, 514, 502          ES\
955      502 DO 510 I = 1, N                 ES\
956      I1 = I - 1                          ES\
957      IF(I1) 503, 505, 503               ES\
958      503 DO 504 J = 1, I1                 ES\
959      504 VI(I) = VI(I) - B(J,I)*VI(J)    ES\
960      IF(ICC) 505, 506, 505               ES\
961      505 IF(B(I,I)) 506, 507, 506        ES\
962      506 VI(I) = VI(I)/B(I,I)            ES\
963      GO TO 510                           ES\
964      507 IF(VI(I)) 508, 509, 508         ES\
965      508 VI(I) = VI(I)*1.0E+15           ES\
966      GO TO 510                           ES\
967      509 VI(I) = 1.0                     ES\
968      510 CONTINUE                        ES\
969      IF(IVC2) 514, 525, 514              ES\
970      514 DO 522 I = 1, N                 ES\
971      IR = NP1 - I                        ES\
972      IF(I - 1) 515, 517, 515            ES\
973      515 I2 = IR + 1                     ES\
974      DO 516 J = I2, N                    ES\
975      516 XI(IR) = XI(IR) - B(IR,J)*XI(J) ES\
976      IF(ICC) 517, 518, 517               ES\
977      517 IF(B(IR,IR)) 518, 519, 518      ES\
978      518 XI(IR) = XI(IR)/B(IR,IR)        ES\

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979      GO TO 522
980      519 IF(XI(IR)) 520, 521, 520
981      520 XI(IR) = XI(IR)*1.0E+15
982      GO TO 522
983      521 XI(IR)=1.000
984      522 CCNTINUE
985      IF(IVC1) 525, 529, 525
986      525 DO 526 I = 2, N
987          IR = NP1 - I
988          I2 = IR + 1
989          DO 526 J = I2, N
990      526 VI(IR) = VI(IR) - B(J,IR)*VI(J)
991          DO 527 L = 1, N
992          I2 = JRGW(L,1)
993      527 VR(I2) = VI(L)
994          DO 528 L = 1, N
995      528 VI(L) = VR(L)
996      529 IF(RCOTI) 615, 656, 615
      C
      C      FACTOR MATRIX.
      C
997      700 ICC = 0
998      SW1=1.0072
999      DO 701 LL = 1, N
1000      701 IROW(LL,1) = LL
1001      DO 708 K = 1, N1
1002      AMAX=DABS(B(K,K))
1003      IMAX = K
1004      K1 = K + 1
1005      DO 702 I = K1, N
1006      IF(AMAX.GT.CABS(B(I,K))) GO TO 702
1007      AMAX=DABS(B(I,K))
1008      IMAX = I
1009      702 CONTINUE
1010      IF(AMAX.LT. SW1) SW1 = AMAX
1011      IF(AMAX.GE.1.0D-25) GO TO 723
1012      B(K,K)=0.000
1013      ICC = ICC + 1
1014      GO TO 708
1015      723 IF(IMAX.EQ. K) GO TO 704
1016      DO 703 J = 1, N
1017      AMAX = B(K,J)
1018      B(K,J) = B(IMAX,J)
1019      703 B(IMAX,J) = AMAX
1020      I2 = IROW(K,1)
1021      IROW(K,1) = IROW(IMAX,1)
1022      IROW(IMAX,1) = I2
1023      704 DO 707 I = K1, N
1024      IF(B(I,K)) 705, 707, 705
1025      705 B(I,K) = B(I,K)/B(K,K)
1026      DO 706 J = K1, N
1027      706 B(I,J) = B(I,J) - B(K,J)*B(I,K)
1028      707 CONTINUE
1029      708 CONTINUE
1030      AMAX=DABS(B(N,N))

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```

1031      IF(AMAX-1.00-25) 712,712,713
1032      712  B(N,N)=0.000
1033      SW1=C.000
1034      ICC = ICC + 1
1035      GO TO 709
1036      713  IF(AMAX .LT. SW1) SW1 = AMAX
1037      709  IF(ICC .LE. ISW) GO TO 710
1038      IF(MP) 1050,1050,1051
1039      1051  WRITE(103,102) ICC
1040      COUNT = C
1041      RETURN
1042      1050  WRITE(103,1052) ICC
1043      710  DO 711 LL = 1, N
1044      711  I2 = IRCH (LL,1)
1045      IROW(I2,2) = LL
1046      IF(RCOTI) 607, 652, 607
1047      1052  FORMAT(///23H ***** WARNING ***** , ' SUBROUTINE EIGVEC HAS
           1 FOUND AN EIGENVALUE CF APPARENT MULTIPLICITY',
           14,/,23X, ' COMPUTATION OF EIGENVECTORS
           2 GENVECTOR(S) CONTINUES AT USER S OPTION'//)
1048      101  FORMAT(38HMORE THAN 15 LCCPS FOR EIGENVECTOR OF,2E12.4,
           2 14H DIFFERENCE CF,E12.4)
1049      102  FORMAT(16H*****WARNING****, 14, 71H ZERCS ON DIAGONAL OF FACTOREDES
           1 MATRIX. CHECK FOR MULTIPLE EIGENVALUES./20X,
           2' SUPROUTINE EIGVEC WILL NOT PERFORM COMPUTATION FOR THIS EIGENVECS)
           3TOR '///)
1050      END

```

/DATA